

# Numpy

with his little helper matplotlib

# Numpy

A fundamental science computing package.

A manipulator for high-dimensional data: *ndarray*

# NDArray

*ndarray*: An array with arbitrary dimension and size.

```
In [1]: import numpy as np
```

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

```
In [3]: print(arr)
```

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

# NDArray – Data Types & Shapes

Numpy *ndarray* are associated with **shapes** and **data types (dtype)**

```
In [1]: import numpy as np
```

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

```
In [3]: print(arr)
```

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

```
In [6]: arr.shape
```

```
Out[6]: (3, 3)
```

```
In [7]: arr.dtype
```

```
Out[7]: dtype('int64')
```

Unlike Python list:

- the shape (size) of the *ndarray* is fixed
- all items in *ndarray* should be of the same type (int / float / ...)

# NDAarray – Data Types & Shapes

- `numpy.shape`
  - Shape of *ndarray*
- `numpy.size`
  - Number of elements in *ndarray*
- `numpy.ndim`
  - Shape of shape of *ndarray*

```
In [11]: arr
Out[11]:
array([[1, 2, 3],
       [4, 5, 6],
       [7, 8, 9]])
```

```
In [12]: arr.ndim
Out[12]: 2
```

```
In [13]: arr.shape
Out[13]: (3, 3)
```

```
In [14]: arr.size
Out[14]: 9
```

# NDArray – Data Types & Shapes

(Suggested) Supported data types:

<code>numpy.int8</code>	<code>numpy.uint8</code>	<code>numpy.float16</code>
<code>numpy.int16</code>	<code>numpy.uint16</code>	<code>numpy.float32</code>
<code>numpy.int32</code>	<code>numpy.uint32</code>	<code>numpy.float64</code>
<code>numpy.int64</code>	<code>numpy.uint64</code>	

The same as in C (<stdint.h>)

Default Data Type

# NDAarray – Creation

- `numpy.array`

```
In [15]: arr = np.array([[1,2,3],[4,5,6],[7,8,9]])
```

```
In [16]: arr
```

```
Out[16]:
```

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

```
In [18]: arr = np.array([[1,2,3],[4,5,6],[7,8,9.]])
```

```
In [19]: arr
```

```
Out[19]:
```

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

```
In [20]: arr.dtype
```

```
Out[20]: dtype('float64')
```

# NDAarray – Creation

- `numpy.arange`

- `numpy.arange([start, ]stop, [step, ]dtype=None, *, like=None)`
- Default step size is 1

- `numpy.linspace`

- `numpy.linspace(start, stop, num=50, endpoint=True, retstep=False, dtype=None, axis=0)`

```
In [118]: np.arange(0, 100)
```

```
Out[118]:
```

```
array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15, 16,
        17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33,
        34, 35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49, 50,
        51, 52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63, 64, 65, 66, 67,
        68, 69, 70, 71, 72, 73, 74, 75, 76, 77, 78, 79, 80, 81, 82, 83, 84,
        85, 86, 87, 88, 89, 90, 91, 92, 93, 94, 95, 96, 97, 98, 99])
```

```
In [117]: np.linspace(0, 100)
```

```
Out[117]:
```

```
array([ 0.          ,  2.04081633,  4.08163265,  6.12244898,
         8.16326531, 10.20408163, 12.24489796, 14.28571429,
        16.32653061, 18.36734694, 20.40816327, 22.44897959,
        24.48979592, 26.53061224, 28.57142857, 30.6122449 ,
        32.65306122, 34.69387755, 36.73469388, 38.7755102 ,
        40.81632653, 42.85714286, 44.89795918, 46.93877551,
        48.97959184, 51.02040816, 53.06122449, 55.10204082,
        57.14285714, 59.18367347, 61.2244898 , 63.26530612,
        65.30612245, 67.34693878, 69.3877551 , 71.42857143,
        73.46938776, 75.51020408, 77.55102041, 79.59183673,
        81.63265306, 83.67346939, 85.71428571, 87.75510204,
        89.79591837, 91.83673469, 93.87755102, 95.91836735,
        97.95918367, 100.          ])
```



# NDAarray – Indexing

Single element indexing

C-order indexing

```
In [23]: arr
```

```
Out[23]:
```

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

```
In [24]: arr[2]
```

```
Out[24]: array([7, 8, 9])
```

```
In [25]: arr[2][1]
```

```
Out[25]: 8
```

```
In [26]: arr[2, 1]
```

```
Out[26]: 8
```

# NDAarray – Indexing

## Single element indexing

Out of bounds

```
In [23]: arr
Out[23]:
array([[1, 2, 3],
       [4, 5, 6],
       [7, 8, 9]])
```

```
In [27]: arr[3]
```

-----

```
IndexError
```

```
Cell In[27], line 1
```

```
----> 1 arr[3]
```

Traceback (most recent call last)

```
IndexError: index 3 is out of bounds for axis 0 with size 3
```

# NDAarray – Indexing

## Single element indexing

Negative indices

```
In [31]: arr[-1]
Out[31]: array([7, 8, 9])
```

```
In [32]: arr[-1, -1]
Out[32]: 9
```

---

```
In [23]: arr
Out[23]:
array([[1, 2, 3],
       [4, 5, 6],
       [7, 8, 9]])
```

```
In [24]: arr[2]
Out[24]: array([7, 8, 9])
```

```
In [25]: arr[2][1]
Out[25]: 8
```

```
In [26]: arr[2, 1]
Out[26]: 8
```

# NDAarray – Indexing

## Single element indexing

With tuple indices

```
In [44]: arr
```

```
Out[44]:
```

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

```
In [45]: indices = (0, 0)
```

```
In [46]: arr[indices]
```

```
Out[46]: 1
```

```
In [23]: arr
```

```
Out[23]:
```

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

```
In [24]: arr[2]
```

```
Out[24]: array([7, 8, 9])
```

```
In [25]: arr[2][1]
```

```
Out[25]: 8
```

```
In [26]: arr[2, 1]
```

```
Out[26]: 8
```

# NDAarray – Indexing

## Single element indexing

With list indices? 😞

```
In [47]: indices = [0, 0]
```

```
In [48]: arr[indices]
```

```
Out[48]:
```

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

---

```
In [23]: arr
```

```
Out[23]:
```

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

```
In [24]: arr[2]
```

```
Out[24]: array([7, 8, 9])
```

```
In [25]: arr[2][1]
```

```
Out[25]: 8
```

```
In [26]: arr[2, 1]
```

```
Out[26]: 8
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

- `start`: included
- `stop`: excluded
  
- No bound restrictions 😊

```
In [58]: arr
```

```
Out[58]:
```

```
array([[ 0,  1,  2,  3,  4],  
       [ 5,  6,  7,  8,  9],  
       [10, 11, 12, 13, 14],  
       [15, 16, 17, 18, 19],  
       [20, 21, 22, 23, 24]])
```

```
In [51]: arr[1:7:2]
```

```
Out[51]:
```

```
array([[ 5,  6,  7,  8,  9],  
       [15, 16, 17, 18, 19]])
```

```
In [52]: arr[100:100]
```

```
Out[52]: array([], shape=(0, 5), dtype=int64)
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

- `start`: included
- `stop`: excluded
- `step`: enum step & **direction**

```
In [58]: arr
```

```
Out[58]:
```

```
array([[ 0,  1,  2,  3,  4],  
       [ 5,  6,  7,  8,  9],  
       [10, 11, 12, 13, 14],  
       [15, 16, 17, 18, 19],  
       [20, 21, 22, 23, 24]])
```

```
In [64]: arr[0:5:2]
```

```
Out[64]:
```

```
array([[ 0,  1,  2,  3,  4],  
       [10, 11, 12, 13, 14],  
       [20, 21, 22, 23, 24]])
```

```
In [66]: arr[5:0:-2]
```

```
Out[66]:
```

```
array([[20, 21, 22, 23, 24],  
       [10, 11, 12, 13, 14]])
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

Use with multi-dimension indexing

```
In [58]: arr
```

```
Out[58]:
```

```
array([[ 0,  1,  2,  3,  4],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14],
       [15, 16, 17, 18, 19],
       [20, 21, 22, 23, 24]])
```

```
In [61]: arr[:, :2, ::2]
```

```
Out[61]:
```

```
array([[ 0,  2,  4],
       [10, 12, 14],
       [20, 22, 24]])
```



# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

Ignore dimension

```
In [58]: arr
```

```
Out[58]:
```

```
array([[ 0,  1,  2,  3,  4],  
       [ 5,  6,  7,  8,  9],  
       [10, 11, 12, 13, 14],  
       [15, 16, 17, 18, 19],  
       [20, 21, 22, 23, 24]])
```

```
In [67]: arr[:, 1]
```

```
Out[67]: array([ 1,  6, 11, 16, 21])
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

```
In [58]: arr
```

```
Out[58]:
```

```
array([[ 0,  1,  2,  3,  4],  
       [ 5,  6,  7,  8,  9],  
       [10, 11, 12, 13, 14],  
       [15, 16, 17, 18, 19],  
       [20, 21, 22, 23, 24]])
```

Ignore certain dimension

```
In [67]: arr[:, 1]
```

```
Out[67]: array([ 1,  6, 11, 16, 21])
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

Ignore all dimensions

```
In [71]: arr = np.arange(27).reshape(3,3,3)
```

```
In [72]: arr
```

```
Out[72]:
```

```
array([[[ 0,  1,  2],
        [ 3,  4,  5],
        [ 6,  7,  8]],

       [[ 9, 10, 11],
        [12, 13, 14],
        [15, 16, 17]],

       [[18, 19, 20],
        [21, 22, 23],
        [24, 25, 26]])
```

```
In [73]: arr[:, 1]
```

```
Out[73]:
```

```
array([[ 3,  4,  5],
       [12, 13, 14],
       [21, 22, 23]])
```

```
In [75]: arr[:, :, 1]
```

```
Out[75]:
```

```
array([[ 1,  4,  7],
       [10, 13, 16],
       [19, 22, 25]])
```

```
In [74]: arr[:, 1, :]
```

```
Out[74]:
```

```
array([[ 3,  4,  5],
       [12, 13, 14],
       [21, 22, 23]])
```

```
In [76]: arr[..., 1]
```

```
Out[76]:
```

```
array([[ 1,  4,  7],
       [10, 13, 16],
       [19, 22, 25]])
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

```
In [78]: arr
```

```
Out[78]:
```

```
array([[ 0,  1,  2,  3,  4],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14],
       [15, 16, 17, 18, 19],
       [20, 21, 22, 23, 24]])
```

```
In [79]: arr[[0,0,0]]
```

```
Out[79]:
```

```
array([[0, 1, 2, 3, 4],
       [0, 1, 2, 3, 4],
       [0, 1, 2, 3, 4]])
```

```
In [80]: arr[:, [0,0,0]]
```

```
Out[80]:
```

```
array([[ 0,  0,  0],
       [ 5,  5,  5],
       [10, 10, 10],
       [15, 15, 15],
       [20, 20, 20]])
```

```
In [81]: arr[[0,0,0], [1,2,3]]
```

```
Out[81]: array([1, 2, 3])
```

## Advanced Indexing

### 1. Integer array indexing

```
In [97]: arr
```

```
Out[97]:
```

```
array([[0, 0, 0, 0, 0],
       [1, 1, 1, 1, 1],
       [2, 2, 2, 2, 2],
       [3, 3, 3, 3, 3],
       [4, 4, 4, 4, 4]])
```

```
In [96]: arr[arr, arr]
```

```
Out[96]:
```

```
array([[0, 0, 0, 0, 0],
       [1, 1, 1, 1, 1],
       [2, 2, 2, 2, 2],
       [3, 3, 3, 3, 3],
       [4, 4, 4, 4, 4]])
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

## Advanced Indexing

### 2. Boolean array indexing – Selection

```
In [78]: arr
```

```
Out[78]:
```

```
array([[ 0,  1,  2,  3,  4],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14],
       [15, 16, 17, 18, 19],
       [20, 21, 22, 23, 24]])
```

```
In [102]: arr[[True, True, True, False, True]]
```

```
Out[102]:
```

```
array([[ 0,  1,  2,  3,  4],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14],
       [20, 21, 22, 23, 24]])
```

```
In [103]: arr[:, [True, True, True, False, True]]
```

```
Out[103]:
```

```
array([[ 0,  1,  2,  4],
       [ 5,  6,  7,  9],
       [10, 11, 12, 14],
       [15, 16, 17, 19],
       [20, 21, 22, 24]])
```

# NDAarray – Indexing

## Slice Indexing

`ndarray[start:stop:step]`

**Advanced Indexing creates copies!**

# NDAarray – Editing

## Edit with slice Indexing

```
In [78]: arr
Out[78]:
array([[ 0,  1,  2,  3,  4],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14],
       [15, 16, 17, 18, 19],
       [20, 21, 22, 23, 24]])
```

```
In [154]: arr = 0
```

```
In [155]: arr
Out[155]: 0
```

```
In [137]: arr[:] = 0
```

```
In [150]: arr[0] = 0
```

```
In [152]: arr[0] = [5,4,3,2,1]
```

```
In [138]: arr
```

```
Out[138]:
```

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

```
In [151]: arr
```

```
Out[151]:
```

```
array([[ 0,  0,  0,  0,  0],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14],
       [15, 16, 17, 18, 19],
       [20, 21, 22, 23, 24]])
```

```
In [153]: arr
```

```
Out[153]:
```

```
array([[ 5,  4,  3,  2,  1],
       [ 5,  6,  7,  8,  9],
       [10, 11, 12, 13, 14],
       [15, 16, 17, 18, 19],
       [20, 21, 22, 23, 24]])
```

# NDArray – Broadcasting

```
In [182]: a = np.array([1,2,3])
```

```
In [183]: b = np.array([2])
```

```
In [184]: a * b
```

```
Out[184]: array([2, 4, 6])
```



# NDAarray – Broadcasting

```
In [182]: a = np.array([1, 2, 3])
```

```
In [185]: a.shape
```

```
Out[185]: (3,)
```

```
In [183]: b = np.array([2])
```

```
In [186]: b.shape
```

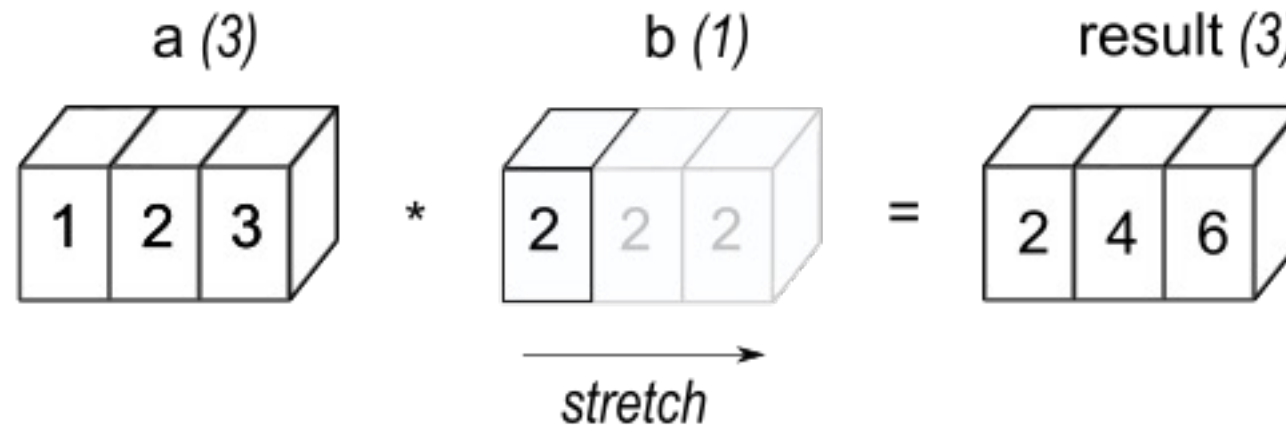
```
Out[186]: (1,)
```

```
In [184]: a * b
```

```
Out[184]: array([2, 4, 6])
```

```
In [187]: (a * b).shape
```

```
Out[187]: (3,)
```



# NDAarray – Broadcasting

Can *ndarrays* be broadcast?

1. Compared the operands from the rightmost dimension
2. Two dimensions are compatible when they are equal, or one of them is 1.

```
A      (4d array):  8 x 1 x 6 x 1
B      (3d array):      7 x 1 x 5
Result (4d array):  8 x 7 x 6 x 5
```

# NDAarray – Basic Functions

`numpy.sort`

```
In [113]: arr
```

```
Out[113]:
```

```
array([[24, 23, 22, 21, 20],  
       [19, 18, 17, 16, 15],  
       [14, 13, 12, 11, 10],  
       [ 9,  8,  7,  6,  5],  
       [ 4,  3,  2,  1,  0]])
```

```
In [114]: np.sort(arr, axis=0)
```

```
Out[114]:
```

```
array([[ 4,  3,  2,  1,  0],  
       [ 9,  8,  7,  6,  5],  
       [14, 13, 12, 11, 10],  
       [19, 18, 17, 16, 15],  
       [24, 23, 22, 21, 20]])
```

```
In [116]: np.sort(arr, axis=1)
```

```
Out[116]:
```

```
array([[20, 21, 22, 23, 24],  
       [15, 16, 17, 18, 19],  
       [10, 11, 12, 13, 14],  
       [ 5,  6,  7,  8,  9],  
       [ 0,  1,  2,  3,  4]])
```

# NDAarray – Basic Functions

`numpy.reshape (numpy.flatten)`

- Reshape *size*

`numpy.transpose`

- Transpose *dimension*

In [163]: arr

Out[163]:

```
array([[ 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 [164]: arr.reshape((3,8))

Out[164]:

```
array([[ 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 [171]: arr

Out[171]:

```
array([[ 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 [172]: arr.transpose((1,0))

Out[172]:

```
array([[ 0,  6, 12, 18],
       [ 1,  7, 13, 19],
       [ 2,  8, 14, 20],
       [ 3,  9, 15, 21],
       [ 4, 10, 16, 22],
       [ 5, 11, 17, 23]])
```

# NDAarray – Basic Functions

## `numpy.squeeze`

- Remove dimension with size 1

## `numpy.newaxis` (== `None`)

- Add a new dimension with size 1

```
In [189]: arr
```

```
Out[189]:
```

```
array([[[[0],  
         [1]],  
       [[2],  
        [3]],  
       [[4],  
        [5]]])
```

```
In [190]: arr.shape
```

```
Out[190]: (3, 2, 1)
```

```
In [191]: np.squeeze(arr)
```

```
Out[191]:
```

```
array([[0, 1],  
       [2, 3],  
       [4, 5]])
```

```
In [192]: np.squeeze(arr).shape
```

```
Out[192]: (3, 2)
```

```
In [194]: np.squeeze(arr)[:, :, np.newaxis]
```

```
Out[194]:
```

```
array([[[[0],  
         [1]],  
       [[2],  
        [3]],  
       [[4],  
        [5]]])
```

```
In [195]: np.squeeze(arr)[:, :, np.newaxis].shape
```

```
Out[195]: (3, 2, 1)
```

# NDArray – Basic Functions

## numpy.stack

- Combine list of arrays, creating new dimension

## numpy.concatenate

- Combine list of arrays, in existing dimensions
- `numpy.hstack`: `partial(numpy.concatenate, axis=0)`
- `numpy.vstack`:
  - For dimension >1: `partial(numpy.concatenate, axis=1)`
  - For dimension =1: `partial(numpy.stack, axis=1)`

```
In [7]: a
Out[7]:
array([[ 0,  1,  2,  3],
       [ 4,  5,  6,  7],
       [ 8,  9, 10, 11]])
```

```
In [8]: np.stack((a,a), axis=0).shape
Out[8]: (2, 3, 4)
```

```
In [9]: np.stack((a,a), axis=1).shape
Out[9]: (3, 2, 4)
```

```
In [10]: np.stack((a,a), axis=2).shape
Out[10]: (3, 4, 2)
```

```
In [11]: np.concatenate((a,a), axis=0).shape
Out[11]: (6, 4)
```

```
In [12]: np.concatenate((a,a), axis=1).shape
Out[12]: (3, 8)
```

```
In [14]: np.vstack((a,a)).shape
Out[14]: (6, 4)
```

```
In [15]: np.hstack((a,a)).shape
Out[15]: (3, 8)
```

# NDArray – Basic Functions

`numpy.sin`

`numpy.cos`

`numpy.tan`

`numpy.sinh`

`numpy.cosh`

`numpy.tanh`

`numpy.arcsin`

`numpy.arccos`

`numpy.arctan`

`numpy.arctan2`

`numpy.exp`

`numpy.exp2`

`numpy.log`

`numpy.log10`

`numpy.log2`

`numpy.add`

`numpy.subtract`

`numpy.multiply`

`numpy.divide`

`numpy.power`

`numpy.min`

`numpy.fmin`

`numpy.max`

`numpy.fmax`

`numpy.sqrt`

`numpy.fabs`

# NDArray – Basic Functions

## linalg: Linear algebra

`numpy.linalg.dot`

`numpy.linalg.multi_dot`

`numpy.linalg.inner`

`numpy.linalg.outer`

`numpy.linalg.matmul (a @ b)`

`numpy.linalg.matrix_power`

`numpy.linalg.qr`

`numpy.linalg.svd`

`numpy.linalg.eigen`

`numpy.linalg.solve`

`numpy.linalg.inv`



# NDArray – Universal Functions (ufunc)

An elegant way to perform batch operations!

`numpy.ufunc.reduce`

`numpy.ufunc.accumulate`

`numpy.ufunc.outer`

`numpy.ufunc.at`

`numpy.ufunc.reduceat`

We use `numpy.add` as example of ufunc!

# NDAarray – Universal Functions (ufunc)

## numpy.ufunc.reduce

In [44]: a

Out[44]:

```
array([[ 0,  1,  2,  3],
       [ 4,  5,  6,  7],
       [ 8,  9, 10, 11]])
```

In [39]: np.add.reduce(a, axis=0)

Out[39]: array([12, 15, 18, 21])

In [40]: np.add.reduce(a, axis=1)

Out[40]: array([ 6, 22, 38])

In [41]: np.add.reduce(a, axis=0, keepdims=True)

Out[41]: array([[12, 15, 18, 21]])

In [42]: np.add.reduce(a, axis=1, keepdims=True)

Out[42]:

```
array([[ 6],
       [22],
       [38]])
```

In [43]: np.add.reduce(a, axis=0, initial=100)

Out[43]: array([112, 115, 118, 121])

# NDAarray – Universal Functions (ufunc)

numpy.ufunc.accumulate

In [44]: a

Out[44]:

```
array([[ 0,  1,  2,  3],
       [ 4,  5,  6,  7],
       [ 8,  9, 10, 11]])
```

In [46]: np.add.accumulate(a, axis=0)

Out[46]:

```
array([[ 0,  1,  2,  3],
       [ 4,  6,  8, 10],
       [12, 15, 18, 21]])
```

In [47]: np.add.accumulate(a, axis=1)

Out[47]:

```
array([[ 0,  1,  3,  6],
       [ 4,  9, 15, 22],
       [ 8, 17, 27, 38]])
```

# NDArray – Universal Functions (ufunc)

numpy.ufunc.outer

In [44]: a

Out[44]:

```
array([[ 0,  1,  2,  3],
       [ 4,  5,  6,  7],
       [ 8,  9, 10, 11]])
```

In [49]: b

Out[49]:

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

In [50]: np.add.outer(a,b)

Out[50]:

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

       [[ 1,  2],
         [ 3,  4]],

       [[ 2,  3],
         [ 4,  5]],

       [[ 3,  4],
         [ 5,  6]]],

      [[[ 4,  5],
         [ 6,  7]],

       [[ 5,  6],
         [ 7,  8]],

       [[ 6,  7],
         [ 8,  9]],

       [[ 7,  8],
         [ 9, 10]]],

      [[[ 8,  9],
         [10, 11]],

       [[ 9, 10],
         [11, 12]],

       [[10, 11],
         [12, 13]],

       [[11, 12],
         [13, 14]]]])
```

In [51]: np.add.outer(a,b).shape

Out[51]: (3, 4, 2, 2)

# NDArray – Universal Functions (ufunc)

numpy.ufunc.at

**In-place operation!**

```
In [44]: a
```

```
Out[44]:
```

```
array([[ 0,  1,  2,  3],  
       [ 4,  5,  6,  7],  
       [ 8,  9, 10, 11]])
```

```
In [76]: np.add.at(a, ([0, 1], [2, 2]), 1)
```

```
In [77]: a
```

```
Out[77]:
```

```
array([[ 0,  1,  3,  3],  
       [ 4,  5,  7,  7],  
       [ 8,  9, 10, 11]])
```

```
In [69]: np.add.at(a, (0, 2), 1)
```

```
In [70]: a
```

```
Out[70]:
```

```
array([[ 0,  1,  3,  3],  
       [ 4,  5,  6,  7],  
       [ 8,  9, 10, 11]])
```

```
In [79]: np.add.at(a, [0, 2], 1)
```

```
In [80]: a
```

```
Out[80]:
```

```
array([[ 1,  2,  3,  4],  
       [ 4,  5,  6,  7],  
       [ 9, 10, 11, 12]])
```

# NDArray – Universal Functions (ufunc)

numpy.ufunc.reduceat

**Reduction operation on a single axis, according to the indices**

```
In [44]: a
```

```
Out[44]:
```

```
array([[ 0,  1,  2,  3],  
       [ 4,  5,  6,  7],  
       [ 8,  9, 10, 11]])
```

```
In [94]: np.add.reduceat(a, [0, 2], axis=1)
```

```
Out[94]:
```

```
array([[ 1,  5],  
       [ 9, 13],  
       [17, 21]])
```

```
In [107]: np.add.reduceat(a, [0, 1, 2, 3], axis=1)
```

```
Out[107]:
```

```
array([[ 0,  1,  2,  3],  
       [ 4,  5,  6,  7],  
       [ 8,  9, 10, 11]])
```