



Numpy



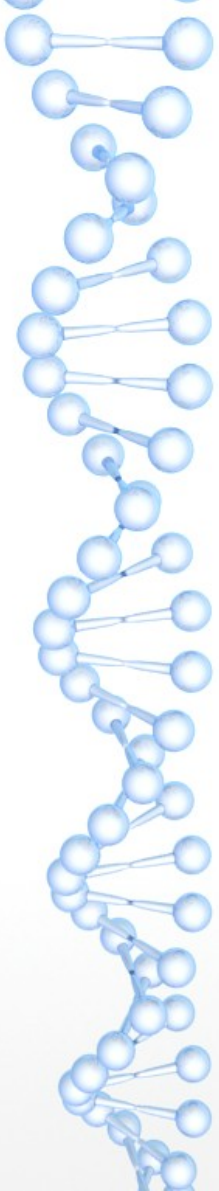
Why Numpy ?

- What is Numpy ?
- Why Numpy ?
- How fast can it be? (Compared to Python)

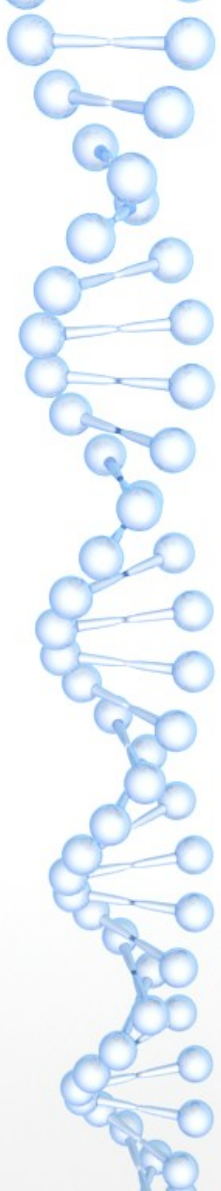


Numerical Python

- ndarray
- functions
- tools
- C, C++, Fortran(API)



Scipy Matplotlib Scikits ...



```
In [7]: 1 import numpy as np
        2 import cupy as cp
        3 np_arr = np.arange(100000000)
        4 cp_arr = cp.arange(100000000)
        5 py_list = list(range(100000000))
```

```
In [8]: 1 %time for _ in range(10): np_arr2 = np_arr * 2
```

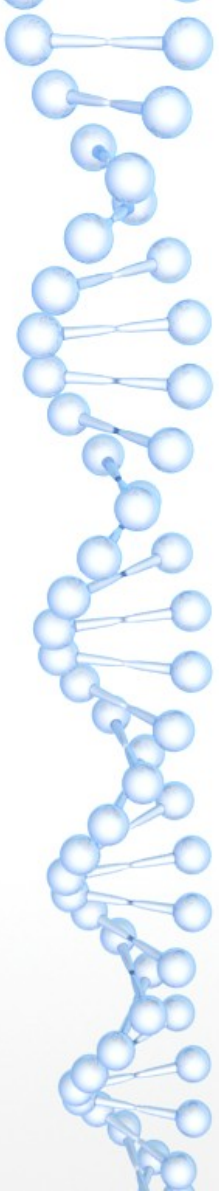
CPU times: user 928 ms, sys: 713 ms, total: 1.64 s
Wall time: 1.64 s

```
In [9]: 1 %time for _ in range(10): py_list2 = [x * 2 for x in py_list]
```

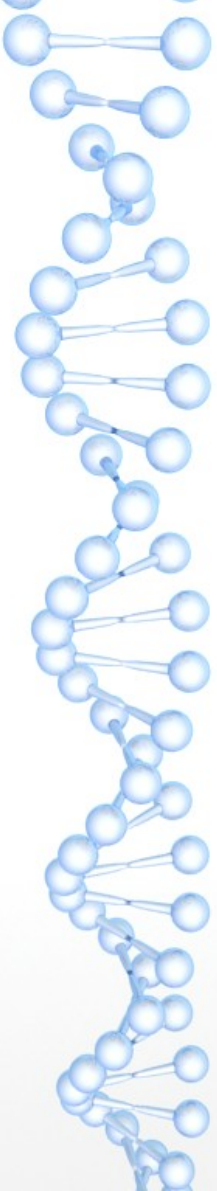
CPU times: user 36.7 s, sys: 7.53 s, total: 44.3 s
Wall time: 44.3 s

```
In [10]: 1 %time for _ in range(10): cp_arr2 = cp_arr * 2
```

CPU times: user 0 ns, sys: 8.08 ms, total: 8.08 ms
Wall time: 7.41 ms



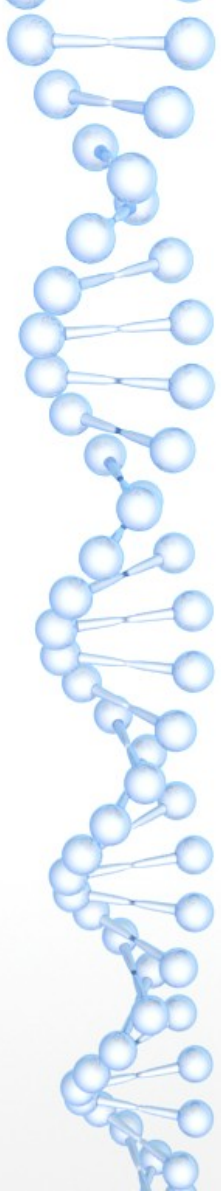
- Interpreted language
- Compiled language



- Dynamically Typed Language
- Statically Typed Language

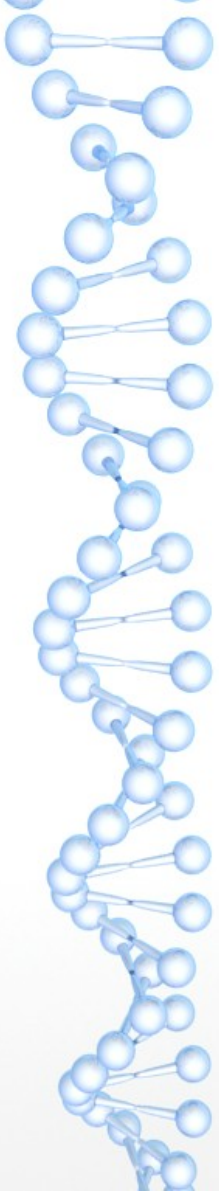


Global Interpreter Lock



Numba/Pypy
Efficiency in time

Cython
Performance
Static variable



Numpy

- C(Compiled language)
- BLAS
- Optimization



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

data

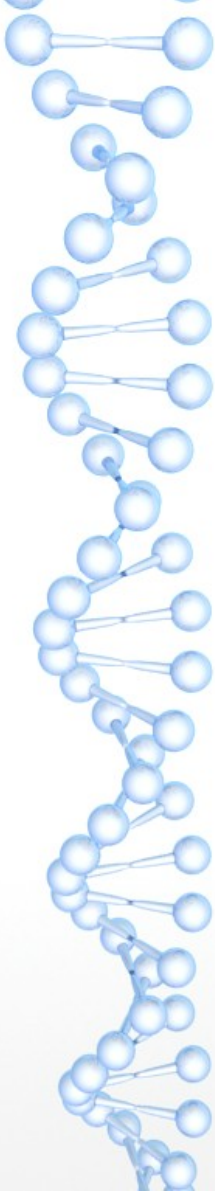
1
2
3

data

1
2
3

.max() =

3



Command

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



NumPy Array

1
2
3



`np.ones(3)`



1
1
1

`np.zeros(3)`

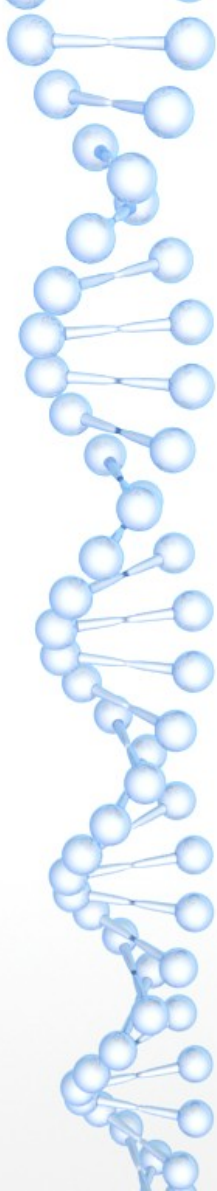


0
0
0

`np.random.random(3)`



0.5967
0.0606
0.2223



data = `np.array([1,2])`

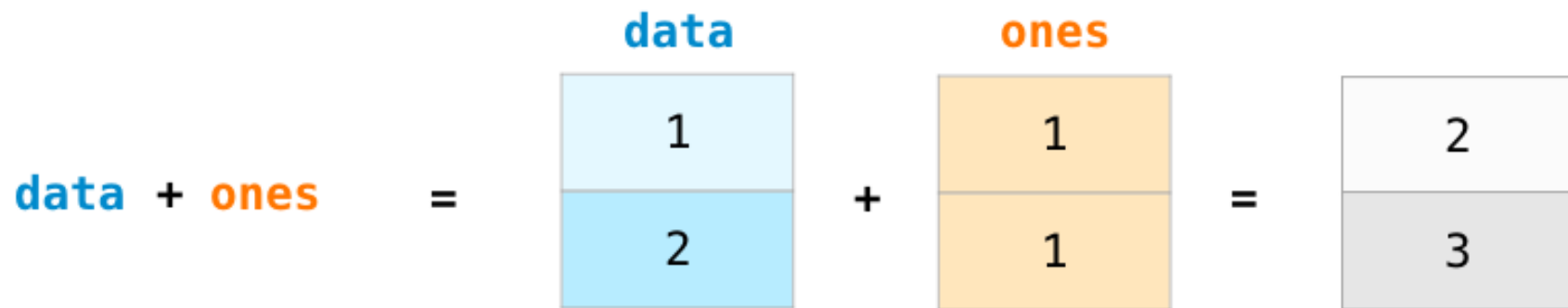
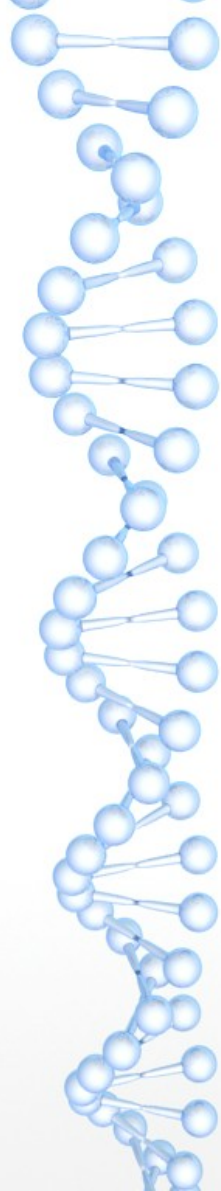
data

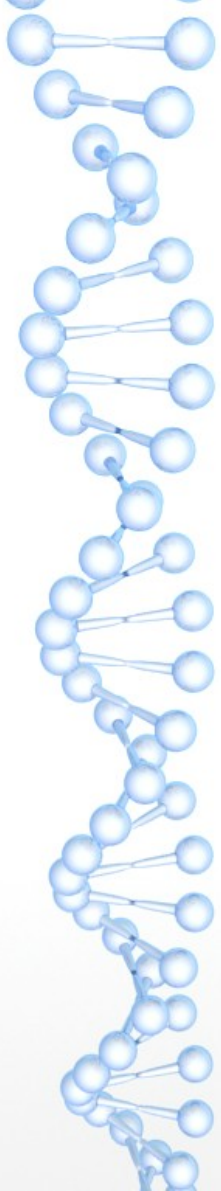
1
2

ones = `np.ones(2)`

ones

1
1





data

1
2

-

ones

1
1

=

0
1

data

1
2

*

data

1
2

=

1
4

data

1
2

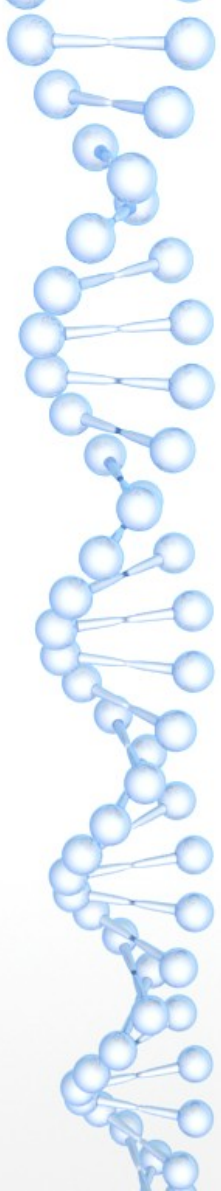
/

data

1
2

=

1
1



1
2

***** **1.6**

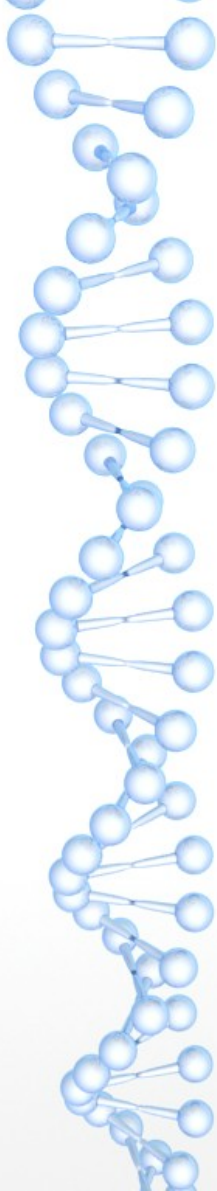
=

1
2

1.6
1.6

=

1.6
3.2



data

0

1

1

2

2

3

data[0]

1

data[1]

2

data[0:2]

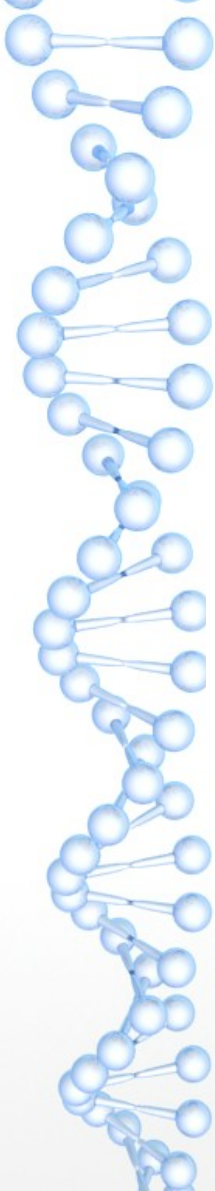
1

2

data[1:]

2

3



data

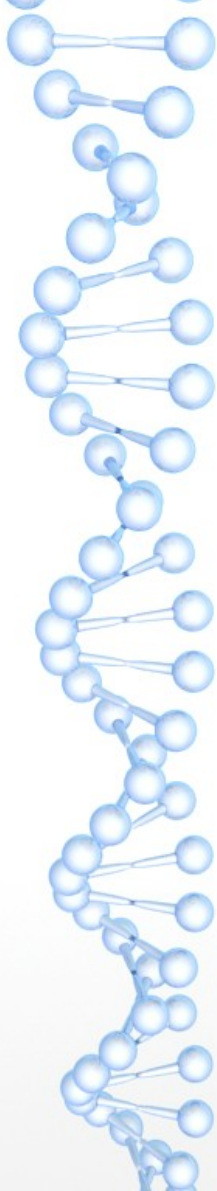
1	
2	<code>.max()</code> =
3	3

data

1	
2	<code>.min()</code> =
3	1

data

1	
2	<code>.sum()</code> =
3	6



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



1	2
3	4



`np.ones((3,2))`

→ 3

2	
1	1
1	1
1	1

`np.zeros((3,2))`

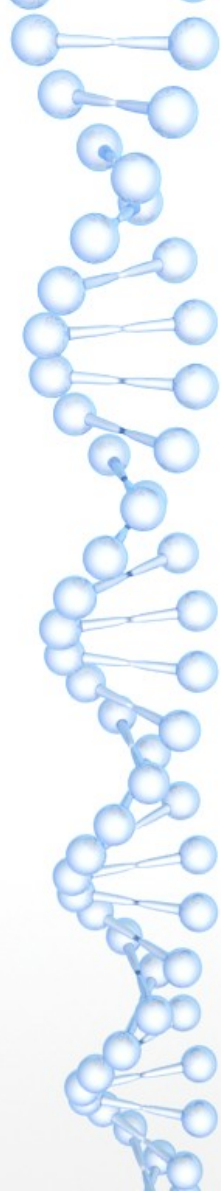
→

0	0
0	0
0	0

`np.random.random((3,2))`

→

0.37	0.88
0.75	0.79
0.63	0.16



data + **ones** =

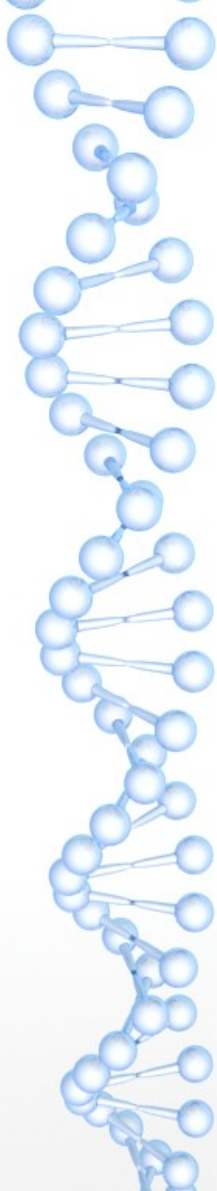
1	2
3	4

+

1	1
1	1

=

2	3
4	5



data + **ones_row** =

data

1	2
3	4
5	6

+

ones_row

1	1
---	---

=

data

1	2
3	4
5	6

+

ones_row

1	1
1	1
1	1

=

2	3
4	5
6	7

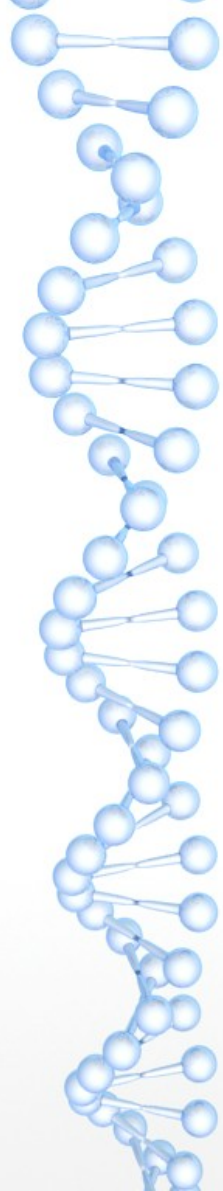
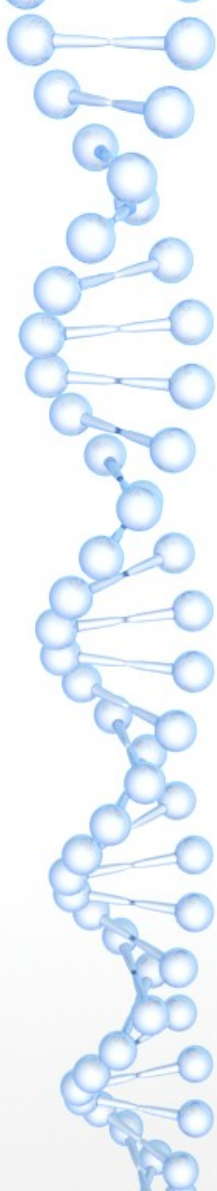


Diagram illustrating matrix multiplication:

data (1x3) \cdot **powers_of_ten** (3x2) = **result** (1x2)

Matrix dimensions: 1x3 3x2 1x2

data			powers_of_ten		result	
1	2	3	1	10	30201	302010
			100	1,000		
			10,000	100,000		



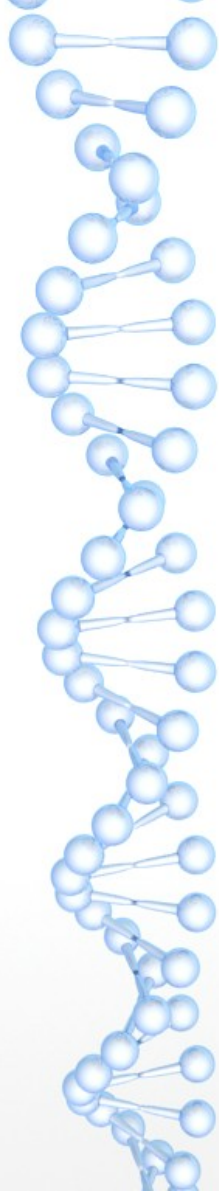
sum(1	100	10,000)	sum(10	1,000	100,000)
	*	*	*			*	*	*	
	1	2	3			1	2	3	

1x2

$1*1 + 2*100 + 3*10,000$	$1*10 + 2*1,000 + 3*100,000$
--------------------------	------------------------------

=

30201	302010
-------	--------



data

	0	1
0	1	2
1	3	4
2	5	6

data[0,1]

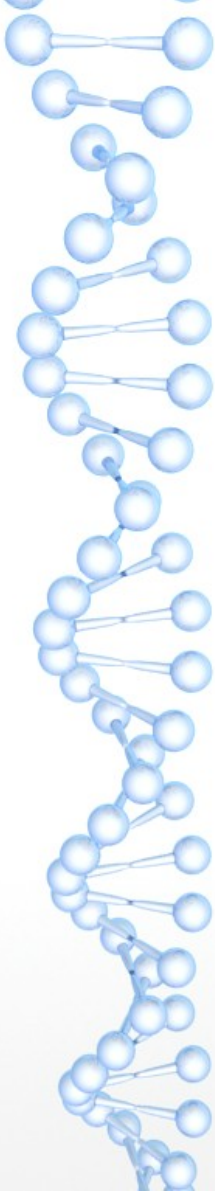
	0	1
0	1	2
1	3	4
2	5	6

data[1:3]

	0	1
0	1	2
1	3	4
2	5	6

data[0:2,0]

	0	1
0	1	2
1	3	4
2	5	6



data

1	2
3	4
5	6

.max() = 6

data

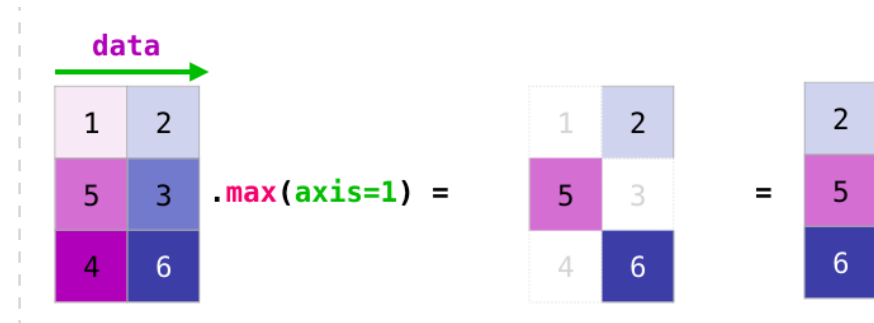
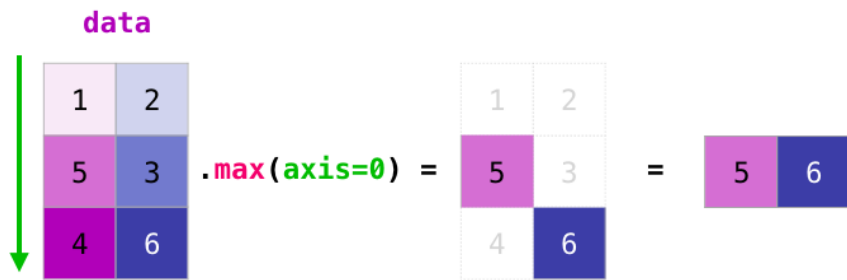
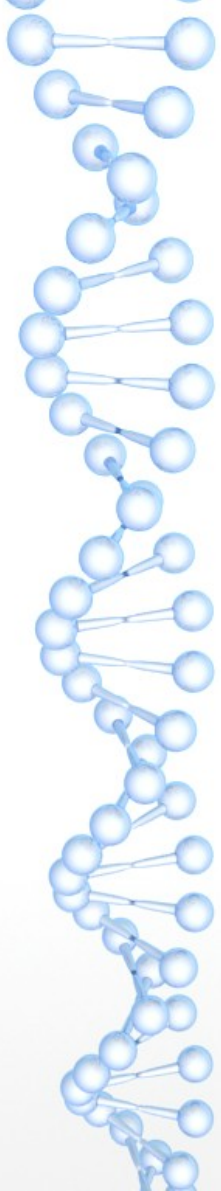
1	2
3	4
5	6

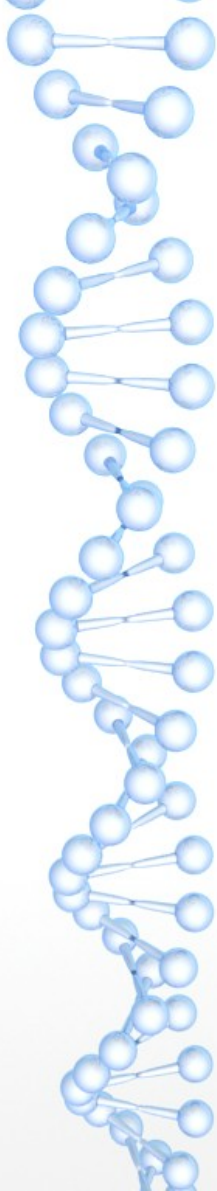
.min() = 1

data

1	2
3	4
5	6

.sum() = 21



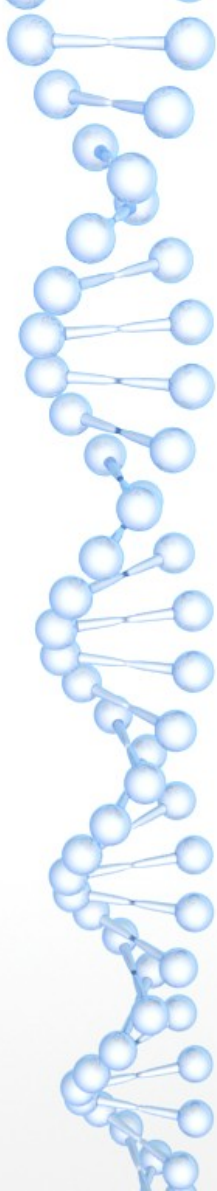


data

1	2
3	4
5	6

data.T

1	3	5
2	4	6



data

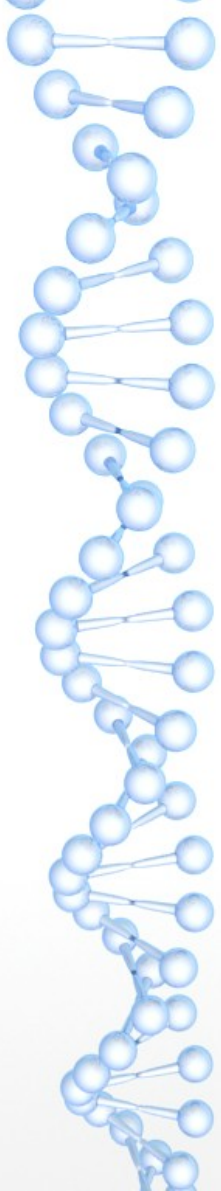
1
2
3
4
5
6

data.reshape(2,3)

1	2	3
4	5	6

data.reshape(3,2)

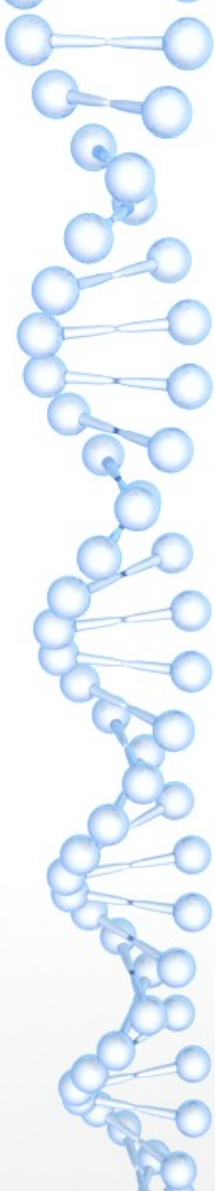
1	2
3	4
5	6



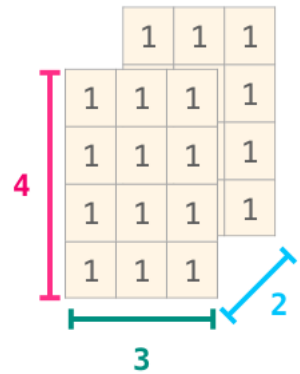
```
np.array([ [[1,2],[3,4]],  
           [[5,6],[7,8]] ])
```



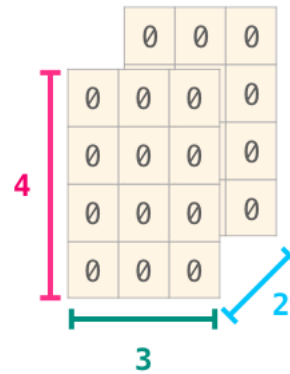
		5	6
1	2		8
3	4		



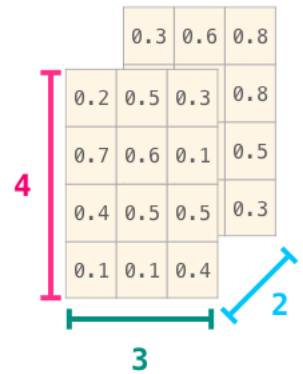
`np.ones((4,3,2))`

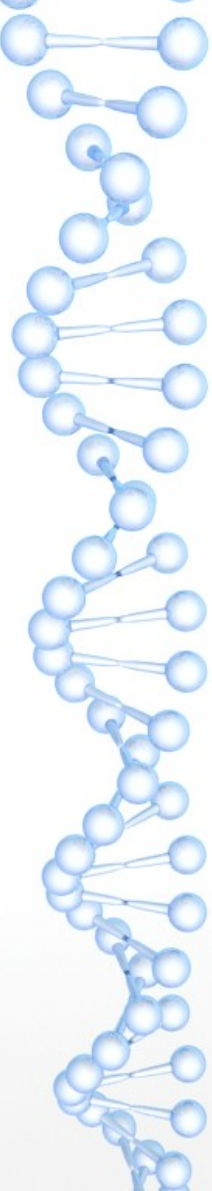


`np.zeros((4,3,2))`



`np.random.random((4,3,2))`





$$MeanSquareError = \frac{1}{n} \sum_{i=1}^n (Y_{prediction_i} - Y_i)^2$$

```
error = (1/n) * np.sum(np.square(predictions - labels))
```

predictions labels

```
error = (1/3) * np.sum(np.square(
```

1
1
1

-

1
2
3

))

```
error = (1/3) * np.sum(np.square(
```

0
-1
-2

))

```
error = (1/3) * np.sum(
```

0
1
4

)

```
error = (1/3) * 5
```

ndarray 对象的属性

属性	说明
<code>.ndim</code>	秩，即轴的数量或维度的数量
<code>.shape</code>	ndarray对象的尺度，对于矩阵，n行m列
<code>.size</code>	ndarray对象元素的个数，相当于 <code>.shape</code> 中 $n*m$ 的值
<code>.dtype</code>	ndarray对象的元素类型
<code>.itemsize</code>	ndarray对象中每个元素的大小，以字节为单位

`.data` 存储地址

Table 4-2. NumPy data types

Type	Type code	Description
int8, uint8	i1, u1	Signed and unsigned 8-bit (1 byte) integer types
int16, uint16	i2, u2	Signed and unsigned 16-bit integer types
int32, uint32	i4, u4	Signed and unsigned 32-bit integer types
int64, uint64	i8, u8	Signed and unsigned 64-bit integer types
float16	f2	Half-precision floating point
float32	f4 or f	Standard single-precision floating point; compatible with C float
float64	f8 or d	Standard double-precision floating point; compatible with C double and Python float object
float128	f16 or g	Extended-precision floating point
complex64, complex128, complex256	c8, c16, c32	Complex numbers represented by two 32, 64, or 128 floats, respectively
bool	?	Boolean type storing True and False values
object	O	Python object type; a value can be any Python object
string_	S	Fixed-length ASCII string type (1 byte per character); for example, to create a string dtype with length 10, use 'S10'
unicode_	U	Fixed-length Unicode type (number of bytes platform specific); same specification semantics as string_ (e.g., 'U10')

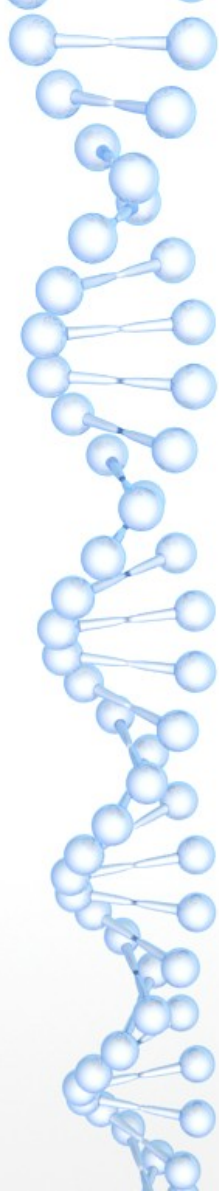


Table 4-3. Unary ufuncs

Function	Description
<code>abs</code> , <code>fabs</code>	Compute the absolute value element-wise for integer, floating-point, or complex values
<code>sqrt</code>	Compute the square root of each element (equivalent to <code>arr ** 0.5</code>)
<code>square</code>	Compute the square of each element (equivalent to <code>arr ** 2</code>)
<code>exp</code>	Compute the exponent e^x of each element
<code>log</code> , <code>log10</code> , <code>log2</code> , <code>log1p</code>	Natural logarithm (base e), log base 10, log base 2, and $\log(1 + x)$, respectively
<code>sign</code>	Compute the sign of each element: 1 (positive), 0 (zero), or -1 (negative)
<code>ceil</code>	Compute the ceiling of each element (i.e., the smallest integer greater than or equal to that number)
<code>floor</code>	Compute the floor of each element (i.e., the largest integer less than or equal to each element)
<code>rint</code>	Round elements to the nearest integer, preserving the dtype
<code>modf</code>	Return fractional and integral parts of array as a separate array
<code>isnan</code>	Return boolean array indicating whether each value is NaN (Not a Number)
<code>isfinite</code> , <code>isinf</code>	Return boolean array indicating whether each element is finite (non- <code>inf</code> , non-NaN) or infinite, respectively
<code>cos</code> , <code>cosh</code> , <code>sin</code> , <code>sinh</code> , <code>tan</code> , <code>tanh</code>	Regular and hyperbolic trigonometric functions
<code>arccos</code> , <code>arccosh</code> , <code>arcsin</code> , <code>arcsinh</code> , <code>arctan</code> , <code>arctanh</code>	Inverse trigonometric functions
<code>logical_not</code>	Compute truth value of <code>not x</code> element-wise (equivalent to <code>~arr</code>).

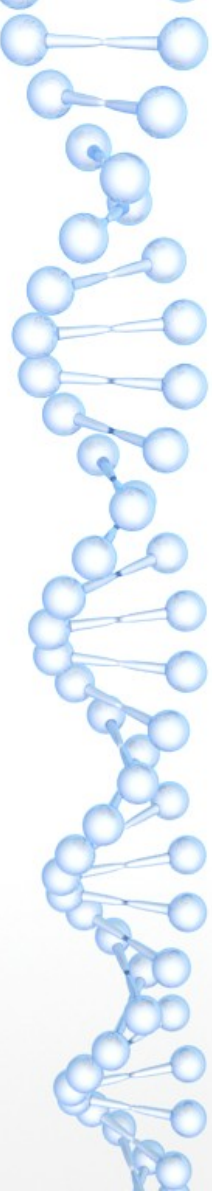
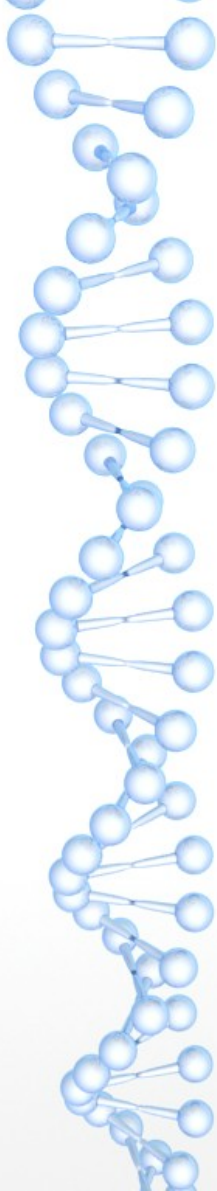


Table 4-4. Binary universal functions

Function	Description
<code>add</code>	Add corresponding elements in arrays
<code>subtract</code>	Subtract elements in second array from first array
<code>multiply</code>	Multiply array elements
<code>divide</code> , <code>floor_divide</code>	Divide or floor divide (truncating the remainder)
<code>power</code>	Raise elements in first array to powers indicated in second array
<code>maximum</code> , <code>fmax</code>	Element-wise maximum; <code>fmax</code> ignores NaN
<code>minimum</code> , <code>fmin</code>	Element-wise minimum; <code>fmin</code> ignores NaN
<code>mod</code>	Element-wise modulus (remainder of division)
<code>copysign</code>	Copy sign of values in second argument to values in first argument



View

	A	B
0	1	2
1	3	4
2	5	6

df1

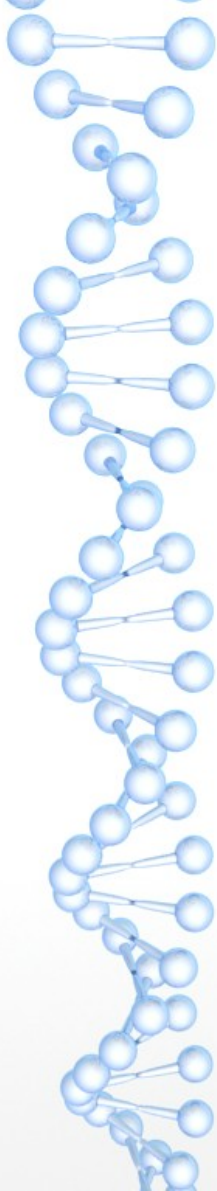
←df2

Copy

	A	B
0	1	2
1	3	4
2	5	6

df1

df2



flatten() 和 ravel() 的区别

数组的扁平化有两种常用的方法，`flatten()` 和 `ravel()`。
。`flatten` 处理后的数组是父数组的引用，因此新数组的任何变化也会改变父数组，因其未用复制的方式构建数组，内存使用效率高，`ravel` 通过复制的方式构建新数组。

