

# Numpy

*Lingua franca for data exchange*

# Numpy

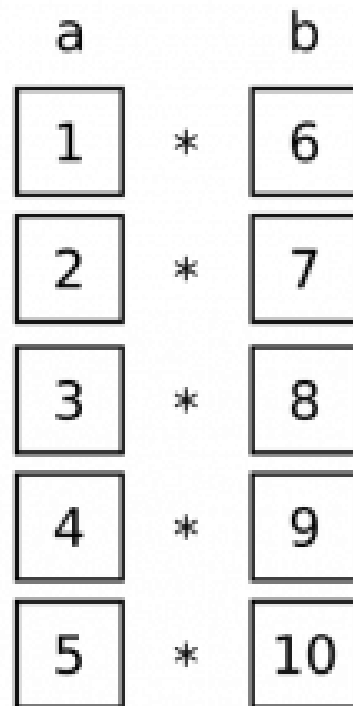
- Numerical Python
- Developed in 2005 by Travis Oliphant
- Lingua franca for data exchange
- ndarray – a n-dimensional array
- Fast operations on entire arrays
- Reading/writing array data
- Linear algebra operations



Travis Oliphant

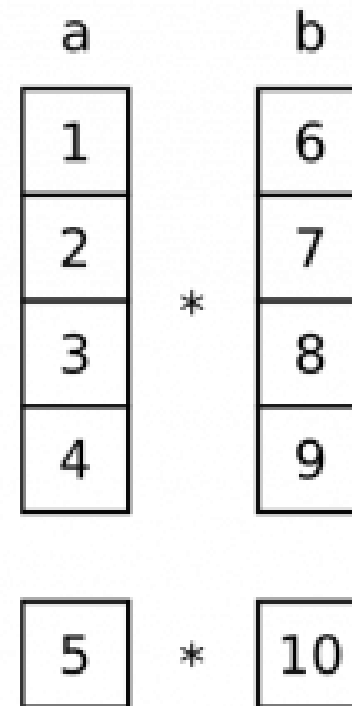
# Vectorized operations

**not vectorized**



5 operations

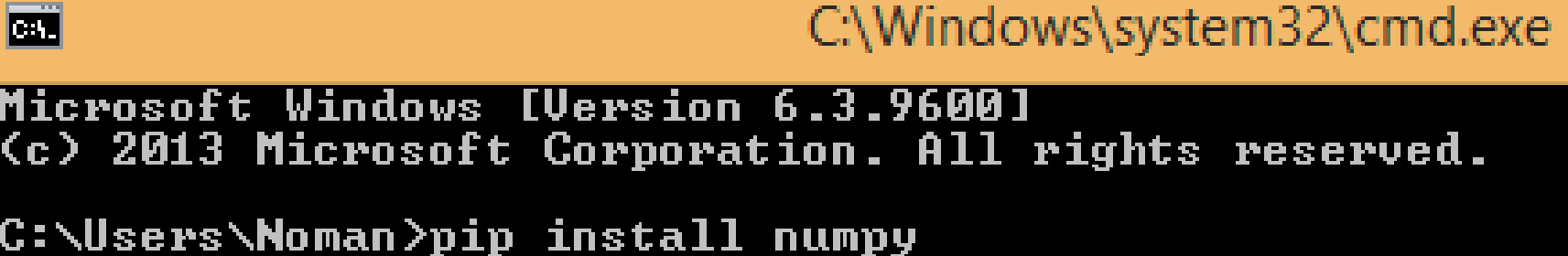
**vectorized**



2 operations

# Installation

- Prepackaged with Anaconda
- Using pip
  - pip install numpy



A screenshot of a Windows command prompt window. The title bar is orange and displays the path `C:\Windows\system32\cmd.exe`. The command prompt icon is visible in the top-left corner. The window content is black with white text. It shows the standard Windows startup message: `Microsoft Windows [Version 6.3.9600]  
(c) 2013 Microsoft Corporation. All rights reserved.` followed by the user's command: `C:\Users\Noman>pip install numpy`.

```
C:\Windows\system32\cmd.exe  
Microsoft Windows [Version 6.3.9600]  
(c) 2013 Microsoft Corporation. All rights reserved.  
C:\Users\Noman>pip install numpy
```

# nd-array

- A fast, flexible container for large datasets in Python
- Homogeneous data i.e. all of the elements must be the same type

# Creating ndarray

- np.array(): convert input data to an ndarray
- np.zeros(): produces arrays of 0s
- np.ones(): produces arrays of 1s
- np.empty(): create new arrays by allocating new memory, but do not populate with any values
- np.arange(): like the built-in range but returns an ndarray instead of a list

# Examples

`np.zeros(9)`

0 0 0 0 0 0 0 0 0

`np.ones(9)`

1 1 1 1 1 1 1 1 1

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

0 0 0  
0 0 0  
0 0 0

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

1 1 1  
1 1 1  
1 1 1

# Attributes of numpy array

- shape: a tuple indicating the size of each dimension
- dtype: an object describing the data type of the array
- ndim: the number of dimensions of the array



# shape attribute

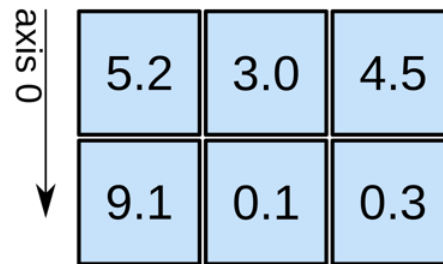
1D array



axis 0 →

shape: (4,)

2D array

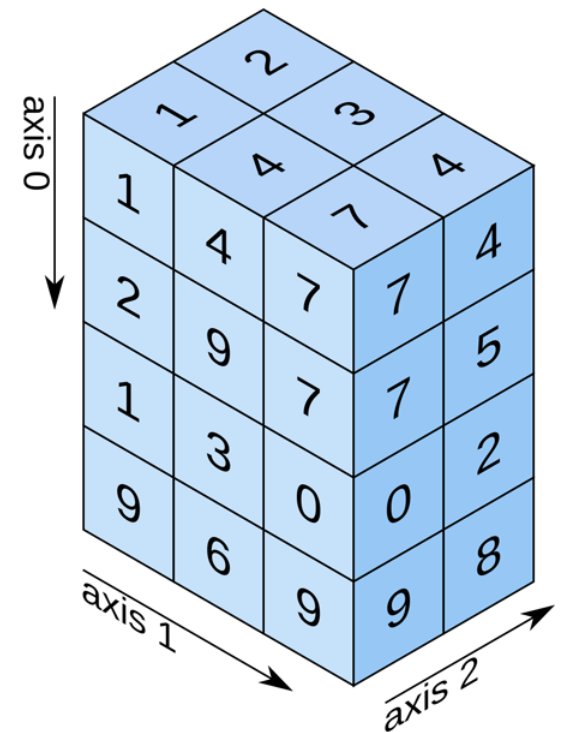


axis 0 ↓

axis 1 →

shape: (2, 3)

3D array



axis 0 ↓

axis 1 →

axis 2 ↗

shape: (4, 3, 2)

# Numpy data types

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

# Vectorization

- Express batch operations on data without writing any for loops
- Any arithmetic operations between equal-size arrays applies the operation element-wise
- Arithmetic operations with scalars propagate the scalar argument to each element in the array
- Comparisons between arrays of the same size yield boolean arrays
- Operations between differently sized arrays is called broadcasting

# Operations on two matrices

```
In [51]: arr = np.array([[1., 2., 3.], [4., 5., 6.]])
```

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

```
Out[52]:
```

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

```
In [53]: arr * arr
```

```
Out[53]:
```

```
array([[ 1.,  4.,  9.],  
       [16., 25., 36.]])
```

```
In [54]: arr - arr
```

```
Out[54]:
```

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

# Operations between matrix and scalar

```
In [55]: 1 / arr
```

```
Out[55]:
```

```
array([[ 1.      ,  0.5     ,  0.3333],  
       [ 0.25    ,  0.2     ,  0.1667]])
```

```
In [56]: arr ** 0.5
```

```
Out[56]:
```

```
array([[ 1.      ,  1.4142,  1.7321],  
       [ 2.      ,  2.2361,  2.4495]])
```

# Comparison between matrices

```
In [4]: a
```

```
Out[4]: array([-0.96206195, -0.98798431, -0.16438853,  1.20690665, -0.56897293,  
              -0.01326404, -0.17575675,  0.76762435,  0.56095689, -0.85087959])
```

```
In [10]: a>0
```

```
Out[10]: array([False, False, False,  True, False, False, False,  True,  True,  
              False])
```

```
In [11]: a[a>0]
```

```
Out[11]: array([1.20690665, 0.76762435, 0.56095689])
```

# Array indexing

```
In [2]: x=np.array([1,7,9,8,5])
```

```
In [3]: x[4]
```

```
Out[3]: 5
```

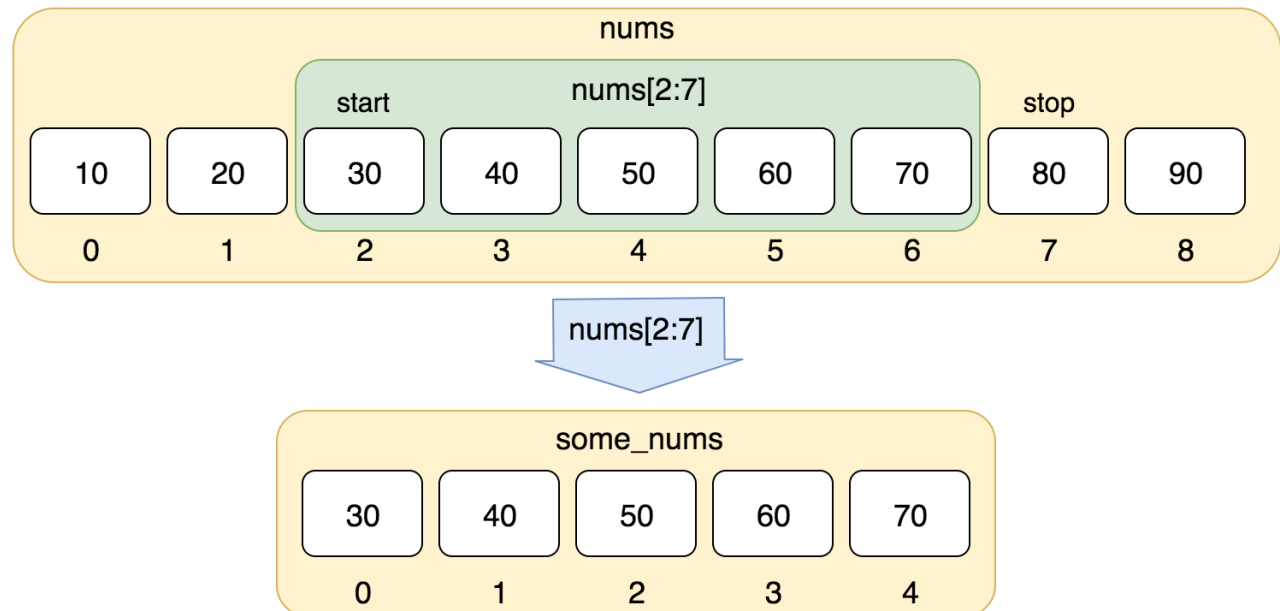
# Fancy indexing

```
In [120]: arr[[4, 3, 0, 6]]
```



# Slicing

- Array slices are views on the original array
- Format
  - start: end: step



# Element-wise array functions

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> ).

# np.where()

`np.where ( condition, option A, option B )`

The diagram illustrates the parameters of the `np.where()` function. It shows the function signature `np.where ( condition, option A, option B )` with three curly braces underneath. The first brace under 'condition' has an arrow pointing down to the text 'The condition'. The second brace under 'option A' has an arrow pointing down to the text 'What to do with entries for which the condition is true'. The third brace under 'option B' has an arrow pointing down to the text 'What to do with entries for which the condition is false'.

The condition

What to do with entries for which the condition is true

What to do with entries for which the condition is false

```
In [3]: salary=np.array([0,-1,100000,50000])
```

```
In [4]: np.where(salary<=0,25000,salary)
```

```
Out[4]: array([ 25000,  25000, 100000,  50000])
```

# Reshape

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

`arr.reshape((4, 3), order=?)`

**C order (row major)**

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

`order='C'`

**Fortran order (column major)**

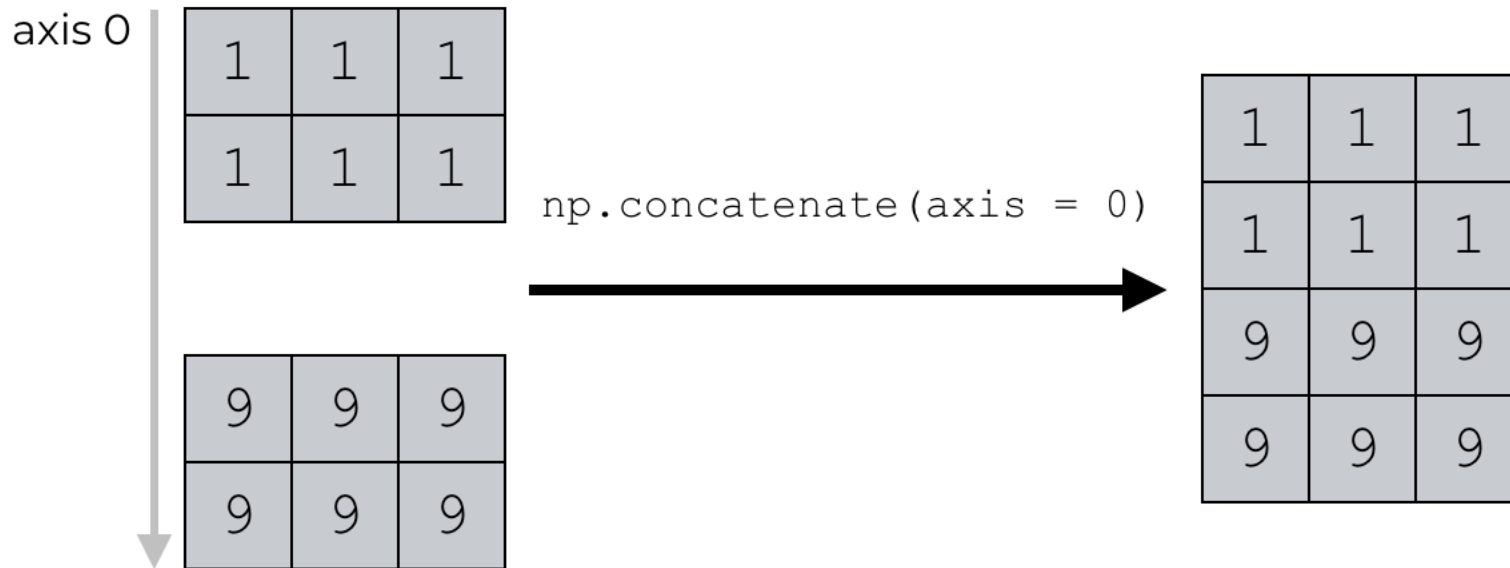
0	4	8
1	5	9
2	6	10
3	7	11

`order='F'`

*Reshaping in C (row major) or Fortran (column major) order*

# concatenate

Setting `axis=0` concatenates along the row axis



# Convenience functions

- vstack: stack arrays row-wise (along axis 0)
- hstack: stack arrays column-wise (along axis 1)

# Splitting an array

