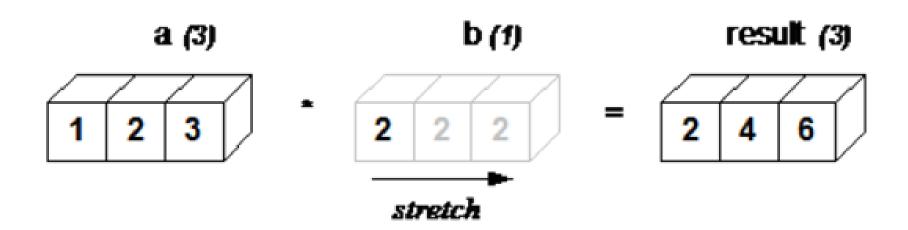
- Broadcasting describes how numpy treats arrays with different shapes during arithmetic operations
- The smaller array is "broadcast" across the larger array so that they have compatible shapes



```
>>> import numpy as np
>>> a = np.array([1, 2, 3])
>>> b = 2
>>> a * b
array([2, 4, 6])
```



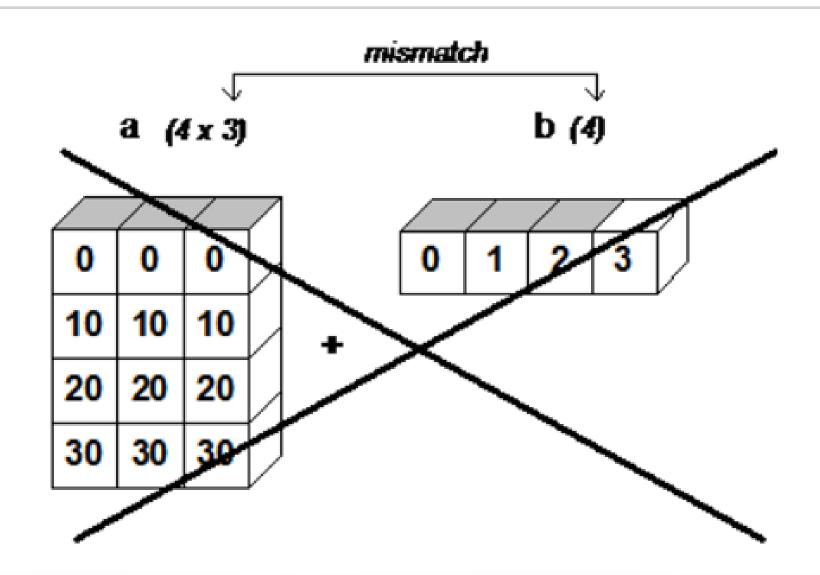
```
>>> a = np.array([[ 0, 0, 0],
                      [10, 10, 10],
                      [20, 20, 20],
                      [30, 30, 30]])
 >>> b = np.array([0, 1, 2])
 >>> a + b
 array([[ 0, 1, 2],
         [10, 11, 12],
         [20, 21, 22],
         [30, 31, 32]])
                                              result (4 x 3)
               a (4 x 3)
                               b (3)
                            0
                10
                                             10
             10
                   10
                                                   12
                20
                                             20
                                                21
                                                   22
             20
                   20
             30
                30 30
                                             30
                            0
                                                31
                                                   32
67
```

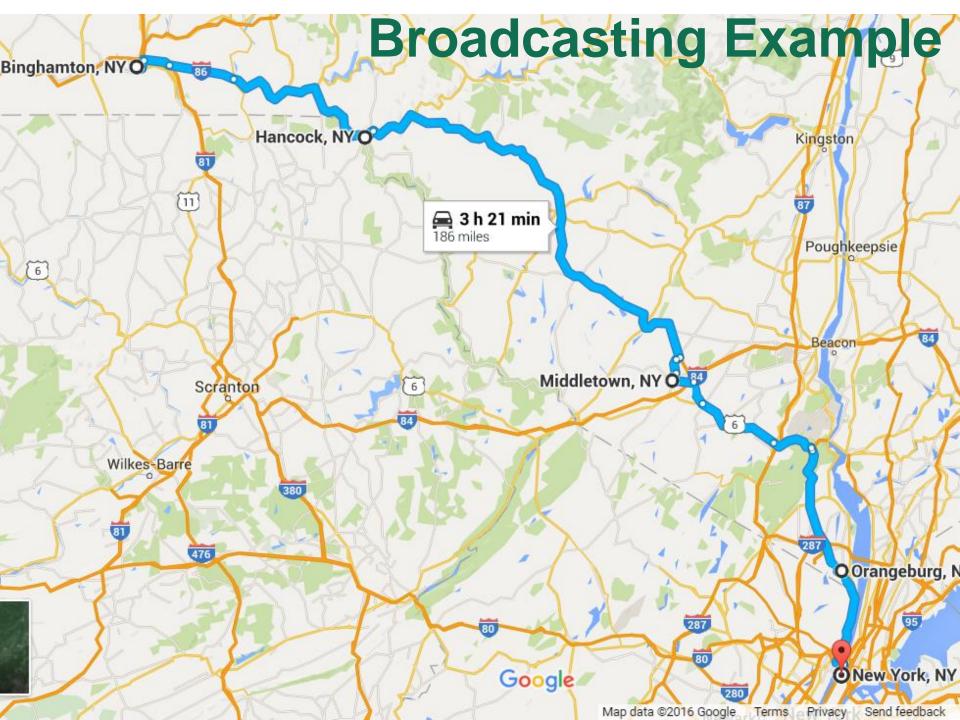
```
>>> a = array([[ 0, 0, 0],
                 [10, 10, 10],
                 [20, 20, 20],
                 [30, 30, 30]])
>>> b = np.array([0,1,2,3]).reshape(4,1)
>>> h
array([[0],
                     10
                        10
                           10
                                             11
                                                11
                                                   11
        [1],
        [2],
                     20
                       20
                          20
                                             22
                                                22
                                                  22
        [3]])
                     30
                       30
                          30
                                             33
                                                33
                                                  33
>>> a + b
                                  stretch
array([[ 0, 0, 0],
        [11, 11, 11],
        [22, 22, 22],
        [33, 33, 33]])
```

```
>>> a = array([[ 0, 0, 0],
                  [10, 10, 10],
                  [20, 20, 20],
                  [30, 30, 30]])
                   0
                      0
                  10
                     10
                         10
                                              20
                                                 20
                                                    20
                             *
                  20
                                             40
                     20
                         20
                                                 40
                                                    40
                                              60
                     30
                         30
                                                 60
                  30
                                                    60
array([[ 0, 0, 0],
                                   stretch
        [20, 20, 20],
        [40, 40, 40],
        [60, 60, 60]])
```

```
>>> a = np.array([0, 10, 20, 30]).reshape(4,1)
>>> b = np.array([0, 1, 2])
>>> a + b
array([[ 0, 1, 2],
        [10, 11, 12],
        [20, 21, 22],
        [30, 31, 32]])
                 a (4x1)
                                                result (4x3)
                               b
                                   (3)
               0
                                   2
                                               0
                                        stretch
                  10
                    10
                                               10
               10
               20
                  20
                             0
                                               20
                                                  21
                                                     22
                    20
               30
                  30
                    30
                                               30
                             0
                                                  31
```

stretch





#### **Broadcasting Example**

Calculate the distances between any two locations

```
(Binghamton, Hancock, Middletown, Orangeburg, NYC)
>>> miles = np.array([0, 43, 120, 168, 186])
>>> miles.reshape(5,1)
array([[ 0],
   [ 43],
      [120],
      [168],
      [186]])
>>> dis array = miles - miles.reshape(5,1)
>>> dis array
array([[ 0, 43, 120, 168, 186],
    [ -43, 0, 77, 125, 143],
      [-120, -77, 0, 48, 66],
      [-168, -125, -48, 0, 18],
      [-186, -143, -66, -18, 0]]
```

#### **NumPy for Matlab Users**



Sdpy.org Docs NumPy v1.11.dev0 Manual

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#### Numpy for Matlab users

#### Introduction

MATLAB® and NumPy/SciPy have a lot in common. But there are many differences. NumPy and SciPy were created to do numerical and scientific computing in the most natural way with Python, not to be MATLAB® clones. This page is intended to be a place to collect wisdom about the differences, mostly for the purpose of heiping. proficient MATLAB® users become proficient NumPy and SciPy users.

#### Some Key Differences

In MATLAB®, the basic data type is a multidimensional array of double. precision floating point numbers. Most expressions take such arrays and return such arrays. Operations on the 2-D instances of these arrays are designed to act more or less like matrix. operations in linear algebra.

MATLAB® uses 1 (one) based. indexing. The initial element of a sequence is found using a(1). See note: INDEXING

MATLABWs scripting language was created for doing linear algebra. The syntax for basic matrix operations is: nice and clean, but the AlPI for adding GUIs and making full-fledged. applications is more or less an afterthought.

In MATLABO, arrays have pass-byvalue semantics, with a lazy copy on write scheme to prevent actually. creating copies until they are actually needed. Slice operations copy parts of the array.

In NumPy the basic type is a multidimensional array. Operations on these arrays in all dimensionalities. including 2D are elementwise operations. However, there is a special matrix type for doing linear algebra, which is just a subclass of the lannay ciasa. Operations on matrix-class. arrays are linear algebra operations. Python uses 0 (zero) based indexing The initial element of a sequence is found using a|0|.

NumPy is based on Python, which was designed from the outset to be an expellent general-purpose programming language. While Maflab's syntax for some array manipulations is more compact than NumPy's, NumPy (by virtue of being an add-on to Python). can do many things that Matlab just cannot, for instance subclassing the main array type to do both array and matrix math cleanly.

In NumPy arrays have pass-byreference semantics. Silce operations are views into an array.

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# Performance Comparison Between Python and NumPy

