

# Multi-Dimensional Arrays

- Rules for arrays of any number of dimensions:

1. Numbers directly separated by commas are in separate columns,
2. Innermost bracketed groups separated by commas are in different rows,
3. Second-innermost bracketed groups separated by commas are on different “pages”,
4. and so on.... zero commas, one number → one column

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

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

one comma, two groups → two rows

one comma, two groups → two pages

... pages, rows, columns

`c.shape`  
`c.ndim`

```
(2L, 2L, 1L)  
3
```

What is the relationship between `c.ndim` and `c.shape`? (Write a Python expression involving them that is always True)

# Indexing Multi-Dimensional Arrays

- Rules for arrays of any number of dimensions:

1. Last index selects the column,
2. Second to last index selects the row,
3. Third to last index selects the page,
4. and so on....

`c = np.random.rand(3)`

`c[1]`

`array([ 0.5664878 , 0.8279805 , 0.04935606])`

`c = np.random.rand(2, 3)`

`c[0,2]`

`c[1,0]`

`c[1, 0, 2]`

`c = np.random.rand(2, 2, 3)`

`array([[[ 0.21282285, 0.85749608, 0.26390172],  
[ 0.55234231, 0.76093889, 0.53984949]],  
[[ 0.95962417, 0.93538958, 0.11380988],  
[ 0.04033667, 0.95009481, 0.77611468]]])`

# Slicing and Fancy Indexing

```
x = np.arange(6)
```

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

How can I use indexing to get:

```
array([0, 5])
```

```
x[::5]  
x[0::5]
```

Slicing

```
x[[0, 5]]  
x[np.array([0, -1])]
```

Fancy Indexing (index list or array)

```
x[x%5==0]  
x[np.logical_or(x<1, x>=5)]
```

Boolean masking

# Copies vs Views

```
x = np.arange(6)
```

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

```
y = x[::5]
```

```
y[0] = 10
```

```
print y
```

```
print x
```

Slicing creates a *view*.

```
array([10, 5])
```

```
array([10, 1, 2, 3, 4, 5])
```

```
y = x[[0,5]]
```

```
y[0] = 10
```

```
print y
```

```
print x
```

Fancy indexing creates a *copy*.

```
array([10, 5])
```

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

# Copies vs Views

```
x = np.arange(10)
```

```
y = np.reshape((2,5))
```

```
y = np.reshape(2,-1)
```

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

```
x[0] = 10
```

```
print y
```

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

# Broadcasting

- Numpy has a rule for performing operations between arrays of different shapes

My attempt at stating it simply:

If dimension  $i$  of array  $a$  is 1 and dimension  $i$  of array  $b$  is  $n$ , then  $a$  is copied along the  $i^{\text{th}}$  dimension  $n$  times.

Now the dimensions of the two arrays match and the operation can be performed.)