

NumPy

Numpy

→ Numpy arrays are of fixed size

Changing the size of ndarray

will create a new array and
delete the original array.

→ Numpy array elements should be
of same datatype.

Why is Numpy Fast ?

- Vectorized code is more concise and easier to read.
- Vectorization results in more "Pythonic" code. Without vectorization, our code would be littered with inefficient and difficult to read for loops.

Basics

→ Numpy object is homogenous
multidimensional
array

Ex: `[[1., 0., 0.],
[0., 1., 2.]]` ndarray

ndarray attributes:

ndarray.ndim

→ No. of dimensions in array

ndarray.shape

→ Dimensions of array

ndarray.size

→ Total no. of elements in array.

`ndarray.dtype`

→ Datatype of elements

`ndarray.itemsize`

→ Size in bytes of each element

`ndarray.data`

→ The buffer containing the actual elements of the array.

Creating a Numpy array

`np.array([2,3,4], dtype=int16)`

`np.zeros((3,4))` # Array full of 0

`np.ones((2,3,4))` # Array full of 1

`np.empty((2,3), dtype=int16)` # Random dimension

→ Default datatype : `float64`

np. arrange (10, 35, 5) # Sequence
 start increment

np. arrange (5, 6, 0.1)

np. linspace (0, 1, 5) # linear spacing
(0., 0.25, 0.5, 0.75, 1.)

Printing Numpy array

print (arr)

Basic Operators

a, b \Rightarrow numpy array

c = a + b

c = a - b

c = a * b

c = a / b

Matrix product

$$\begin{aligned} c = a @ b \\ c = a \cdot \text{dot}(b) \end{aligned} \quad \left. \begin{array}{l} \\ \end{array} \right\} \begin{array}{l} \text{Matrix} \\ \text{cross product} \end{array}$$

Can type cast from

$$\text{int32} \rightarrow \text{float64} \rightarrow \text{complex}$$

but can't type cast from

$$\text{complex} \not\rightarrow \text{float64} \not\rightarrow \text{int32}$$

`np.sum (axis=0)` #Sum of each column

`np.sum (axis=1)` #Sum of each row



`axis=1`

`axis=0`

Indexing , Slicing and Iterating

$a[2]$

$a[2:5]$ start : end

$a[2:5:2]$ start : end : increment

$a[::-1]$ Reversed array

`np.fromfunction (fun, (5, 4), dtype= int)`

$a[2,3]$ # $a[2][3]$

$a[0:5, 1]$ # $a[i][1]$

$a[1, 0:5]$ # $a[1][i]$

$a[1:3, 0:5]$

$b[i, \dots]$

→ ... are used to fill remaining ":"

$b[1, \dots] == b[1, :, :] == b[1]$

$b.flat$

→ Iterator over all the elements of the array.

Change the shape of Numpy

$a.ravel()$ # flattens array

$a.reshape(6,2)$ # changes shape

$a.T$ # Returns transpose of a

$a.resize(2,6)$

$a.resize(3, -1)$

dimension

→ -1 automatically calculates other n

Stacking different arrays

np.vstack(a, b)

$$\begin{bmatrix} a \\ b \end{bmatrix}$$

np.concatenate((a, b), axis=0)

np.hstack(a, b)

$$\begin{bmatrix} a & b \end{bmatrix}$$

np.column_stack(a, b)

$$\begin{bmatrix} [& , &] \\ [& , &] \\ [& , &] \end{bmatrix}$$

np.concatenate((a, b))

$$\begin{matrix} \uparrow \\ a \\ \uparrow \\ b \end{matrix}$$

Splitting Array

np.hsplit(a, 3)

3 equal parts

np.hsplit(a, (3, 4))

Split at column 3 and column 4

View or Shallow Copy

→ Creates a new object that references the main object.

`a.view()`

Deep Copy

→ Creates a new object with new data

`a.copy()`

Indexing with array of indices

```
>>> a = np.arange(12)**2 # the first 12 square numbers
>>> i = np.array([1, 1, 3, 8, 5]) # an array of indices
>>> a[i] # the elements of 'a' at the positions 'i'
array([ 1,  1,  9, 64, 25])
>>>
>>> j = np.array([[3, 4], [9, 7]]) # a bidimensional array of indices
>>> a[j] # the same shape as 'j'
array([[ 9, 16],
       [81, 49]])
```

→ Multi dimensional indexing

```
>>> palette = np.array([[0, 0, 0],           # black
...                      [255, 0, 0],        # red
...                      [0, 255, 0],        # green
...                      [0, 0, 255],        # blue
...                      [255, 255, 255]]) # white
>>> image = np.array([[0, 1, 2, 0], # each value corresponds to a color in the pa
...                      [0, 3, 4, 0]])
>>> palette[image] # the (2, 4, 3) color image
array([[[ 0,  0,  0],
       [255,  0,  0],
       [ 0, 255,  0],
       [ 0,  0,  0]],

       [[ 0,  0,  0],
       [ 0,  0, 255],
       [255, 255, 255],
       [ 0,  0,  0]]])
```

Ex:-

```
>>> a = np.arange(12).reshape(3, 4)
>>> a
array([[ 0,  1,  2,  3],
       [ 4,  5,  6,  7],
       [ 8,  9, 10, 11]])
>>> i = np.array([[0, 1], # indices for the first dim of `a`
...                  [1, 2]])
>>> j = np.array([[2, 1], # indices for the second dim
...                  [3, 3]])
>>>
>>> a[i, j] # i and j must have equal shape
array([[ 2,  5],
       [ 7, 11]])
>>>
>>> a[i, 2]
array([[ 2,  6],
       [ 6, 10]])
>>>
>>> a[:, j]
array([[[ 2,  1],
       [ 3,  3]],

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

       [[10,  9],
       [11, 11]]])
```

np.argmax (axis=0)

→ Returns the index of maximum number in axis 0 (↓)

Indexing with boolean Array

```
>>> a = np.arange(12).reshape(3, 4)
>>> b = a > 4
>>> b # `b` is a boolean with `a`'s shape
array([[False, False, False, False],
       [False, True, True, True],
       [True, True, True, True]])
>>> a[b] # 1d array with the selected elements
array([ 5,  6,  7,  8,  9, 10, 11])
```

```
>>> a[b] = 0 # All elements of `a` higher than 4 become 0
>>> a
array([[0, 1, 2, 3],
       [4, 0, 0, 0],
       [0, 0, 0, 0]])
```

data.min(axis=0) #Min of columns
data.max() #Max value in matrix
data.sum() #Sum of all values
np.unique(data, # Unique values
return_index=True, # Returns index
return_counts=True, # Occurrences
axis=0) #Optional

np.flip(data, # Reverse an array
axis=0) #Optional

Ex: $MSE = \frac{1}{n} \sum_{i=1}^n (Y_{pred_i} - Y_i)^2$

error = $(1/n) * np.sum(np.square(pred - labels))$

Save & Load NumPy File

`np.save ('filename', data)`

`data = np.load ('filename.npy')`

`np.savetxt ('newfile.csv', data)`

`data = np.loadtxt ('newfile.csv')`

`np.nan` \Rightarrow Null value in numpy