

Numpy and Scipy

Numerical Computing in Python

What is Numpy?

- Numpy, Scipy, and Matplotlib provide MATLAB-like functionality in python.
- Numpy Features:
 - Typed multidimensional arrays (matrices)
 - Fast numerical computations (matrix math)
 - High-level math functions

Why do we need NumPy

Let's see for ourselves!

Why do we need NumPy

- Python does numerical computations slowly.
- 1000 x 1000 matrix multiply
 - Python triple loop takes > 10 min.
 - Numpy takes ~0.03 seconds

NumPy Overview

1. Arrays
2. Shaping and transposition
3. Mathematical Operations
4. Indexing and slicing
5. Broadcasting

Arrays

Structured lists of numbers.

- Vectors
- Matrices
- Images
- Tensors
- ConvNets

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$$\begin{bmatrix} p_x \\ p_y \\ p_z \end{bmatrix}$$

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix}$$

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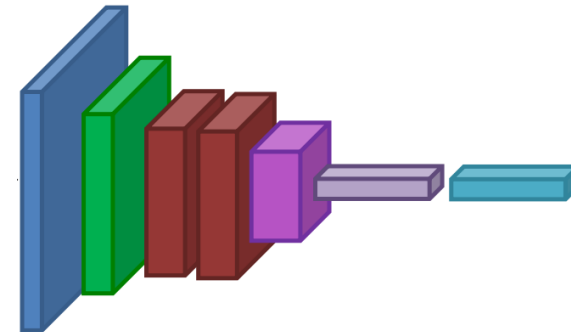
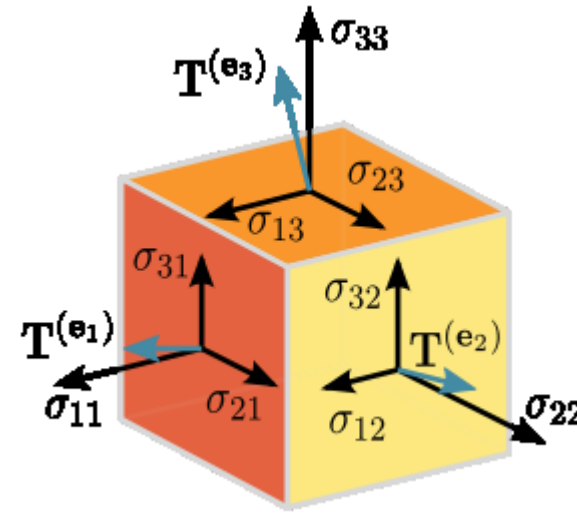
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MATRICES

IMAGES

TENSORS

CONVNETS



Arrays, Basic Properties

```
import numpy as np  
  
a = np.array([[1, 2, 3], [4, 5, 6]], dtype=np.float32)  
  
print(a.ndim, a.shape, a.dtype)
```

1. Arrays can have any number of dimensions, including zero (a scalar).
2. Arrays are typed: `np.uint8`, `np.int64`, `np.float32`, `np.float64`
3. Arrays are dense. Each element of the array exists and has the same type.

Arrays, creation

- `np.ones`, `np.zeros`
- `np.arange`
- `np.concatenate`
- `np.astype`
- `np.zeros_like`,
`np.ones_like`
- `np.random.random`

Arrays, creation

- **np.ones, np.zeros**
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```
>>> np.ones((3,5),dtype=np.float32)
array([[ 1.,  1.,  1.,  1.,  1.],
       [ 1.,  1.,  1.,  1.,  1.],
       [ 1.,  1.,  1.,  1.,  1.]], dtype=float32)
```

```
>>> np.zeros((6,2),dtype=np.int8)
array([[0, 0],
       [0, 0],
       [0, 0],
       [0, 0],
       [0, 0],
       [0, 0]], dtype=int8)
```

Arrays, creation

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```
>>> np.arange(1334,1338)  
array([1334, 1335, 1336, 1337])
```

Arrays, creation

- np.ones, np.zeros
- np.arange
- **np.concatenate**
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np.ones_like
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```
>>> A = np.ones((2,3))
>>> B = np.zeros((4,3))
>>> np.concatenate([A,B])
array([[ 1.,  1.,  1.],
       [ 1.,  1.,  1.],
       [ 0.,  0.,  0.],
       [ 0.,  0.,  0.],
       [ 0.,  0.,  0.],
       [ 0.,  0.,  0.]])
>>>
```

Arrays, creation

- np.ones, np.zeros
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- np.zeros_like,
np.ones_like
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```
>>> A = np.ones((4,1))
>>> B = np.zeros((4,2))
>>> np.concatenate([A,B], axis=1)
array([[ 1.,  0.,  0.],
       [ 1.,  0.,  0.],
       [ 1.,  0.,  0.],
       [ 1.,  0.,  0.]])
```


Arrays, creation

- np.ones, np.zeros
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np.ones_like
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```
>>> A
array([[ 4670.5,  4670.5,  4670.5],
       [ 4670.5,  4670.5,  4670.5],
       [ 4670.5,  4670.5,  4670.5],
       [ 4670.5,  4670.5,  4670.5],
       [ 4670.5,  4670.5,  4670.5]], dtype=float32)
>>> print(A.astype(np.uint16))
[[4670 4670 4670]
 [4670 4670 4670]
 [4670 4670 4670]
 [4670 4670 4670]
 [4670 4670 4670]]
```

Arrays, creation

- np.ones, np.zeros
- np.arange
- np.concatenate
- np.astype
- **np.zeros_like,**
np.ones_like
- np.random.random

```
>>> a = np.ones((2,2,3))  
>>> b = np.zeros_like(a)  
>>> print(b.shape)
```

Arrays, creation

- `np.ones`, `np.zeros`

- `np.arange`

- `np.concatenate`

- `np.astype`

- `np.zeros_like`,
`np.ones_like`

- `np.random.random`

```
>>> np.random.random((10,3))
array([[ 0.61481644,  0.55453657,  0.04320502],
       [ 0.08973085,  0.25959573,  0.27566721],
       [ 0.84375899,  0.2949532 ,  0.29712833],
       [ 0.44564992,  0.37728361,  0.29471536],
       [ 0.71256698,  0.53193976,  0.63061914],
       [ 0.03738061,  0.96497761,  0.01481647],
       [ 0.09924332,  0.73128868,  0.22521644],
       [ 0.94249399,  0.72355378,  0.94034095],
       [ 0.35742243,  0.91085299,  0.15669063],
       [ 0.54259617,  0.85891392,  0.77224443]])
```

Arrays, danger zone

- Must be dense, no holes.
- Must be one type
- Cannot combine arrays of different shape

```
>>> np.ones([7,8]) + np.ones([9,3])  
Traceback (most recent call last):  
  File "<stdin>", line 1, in <module>  
ValueError: operands could not be broadcast together  
with shapes (7,8) (9,3)
```

Shaping

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

```
a = a.reshape(3, 2)
```

```
a = a.reshape(2, -1)
```

```
a = a.ravel()
```

1. Total number of elements cannot change.
2. Use -1 to infer axis shape
3. Row-major by default (MATLAB is column-major)

Return values

- Numpy functions return either **views** or **copies**.
- Views share data with the original array, like references in Java/C++. Altering entries of a view, changes the same entries in the original.
- The [numpy documentation](#) says which functions return views or copies
- `np.copy`, `np.view` make explicit copies and views.

Transposition

```
a = np.arange(10).reshape(5, 2)
```

```
a = a.T
```

```
a = a.transpose((1, 0))
```

`np.transpose` permutes axes.

`a.T` transposes the first two axes.

Saving and loading arrays

```
np.savez('data.npz', a=a)
```

```
data = np.load('data.npz')
```

```
a = data['a']
```

1. NPZ files can hold multiple arrays
2. `np.savez_compressed` similar.