Introduction to Python Day 4

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NumPy

- Python library numerical data
- Very fast (C/C++) and multithreaded)
- Vectorized (substitutes for-loops)
- Good short cut for a lot of things
- Will be your best friend!



NumPy Structures

• similar to what we know already

Basic

- Different data types (\mathtt{dtype}) are valid
- type followed by bit (8, 16, 32, 64, 128)
 - 1. float64
 - 2. int64

- 3. complex64
- 4. bool_
- 5. str_
- 6. object_
- 7. datetime64

Array

- 1D, 2D and nD
 - Matrices are a kind of array with special functions
- Different ways to make them
- Have shape properties

Make filled arrays

np.ones(5)

np.random.rand(5)

• Array with zeros

```
np.zeros(5)
array([0., 0., 0., 0., 0.])
• Arrays with ones
```

```
array([1., 1., 1., 1., 1.])
```

• Arrays filled with random numbers

```
array([0.2337696 , 0.57120003, 0.23800159, 0.85435414, 0.56713309])
```

• Arrays filled with what you want

```
np.full(5, "hello")
array(['hello', 'hello', 'hello', 'hello'], dtype='<U5')</pre>
```

Array from lists

• Convert an existing list to an array

[1, 2, 3]

Array initiation short-cut

• Use the size/shape of another array

```
old_array = np.array([1,2,3,4])
np.zeros_like(old_array)
array([0, 0, 0, 0])
```

n-dimensional arrays

• Every initation of Arrays can be multidimensional

Array with random numbers

- Callback: spike_simulation
- Generate random numbers as arrays

```
np.random.seed(42)
np.random.normal(5, 2, 20)

array([5.99342831, 4.7234714 , 6.29537708, 8.04605971, 4.53169325, 4.53172609, 8.15842563, 6.53486946, 4.06105123, 6.08512009, 4.07316461, 4.06854049, 5.48392454, 1.17343951, 1.55016433, 3.87542494, 2.97433776, 5.62849467, 3.18395185, 2.1753926 ])
```

Changing the shapes of arrays

```
np.random.seed(42)
A = np.random.normal(5, 2, 20)
A = A.reshape(2, 5, 2)
print(A)
print(np.shape(A))
```

```
[[[5.99342831 4.7234714]

[6.29537708 8.04605971]

[4.53169325 4.53172609]

[8.15842563 6.53486946]

[4.06105123 6.08512009]]

[[4.07316461 4.06854049]

[5.48392454 1.17343951]

[1.55016433 3.87542494]

[2.97433776 5.62849467]

[3.18395185 2.1753926]]]

(2, 5, 2)
```

Special cases for arrays

• Arrays can also be "empty"

```
np.empty(10)

array([4.67211910e-310, 0.00000000e+000, 6.89906821e-310, 6.89906820e-310, 6.89906820e-310, 6.89906731e-310, 6.89907586e-310, 6.89907587e-310, 6.89906729e-310, 6.89906821e-310])
```

• Only useful in very specific cases (otherwise danger zone)

Array sequences

• Generate sequences

```
np.arange(start=2, stop=10, step=2)
array([2, 4, 6, 8])
```

- Similar logic to iterators from day 1 (2:10:2)
- Array can also go in steps of floats e.g. 0.2

Array sequences

- Alternative linspace and logspace
- specifying the number of elements we want to have

Exercise time

Indexing

- nonzero()
- where()
- diag()

Special Indices

• Recall finding the maximum

```
# looping through data indices. find the max
B = [1, 4, 6, 7, 89, 54]
big_indx = 0
for i in range(len(B)):
    if B[i] > B[big_indx]:
        big_indx = i
print('The max value in B is', B[big_indx], 'found on position', big_indx)

The max value in B is 89 found on position 4

# looping through data indices. find the max
B = [1, 4, 6, 7, 89, 54]
big_indx = np.argmax(B)
print('The max value in B is', B[big_indx], 'found on position', big_indx)
```

The max value in B is 89 found on position 4

for loops can be often replaced using functions and make your code faster and easier to read. As you can see from the example we can also use a list as function input. Numpy will convert the list automatically, work with an array and return an array too.

Operations

- Lots of useful functions:
 - Mathematical functions
 - Linear algebra
 - Sorting and Counting
 - Statistics
 - Random number generation
 - Input/Output (I/O)
 - Memory mapping (mmap)

Mathematical function

- Vectorized functions
- Versions which handle nan

Statistics

• Get some summary statistics

```
power = np.array([313, 271, 912, 851, 239, 715])
np.mean(power)
np.median(power)
np.std(power)

np.float64(550.166666666666)

np.float64(514.0)

np.float64(282.72508240731355)
```

Functions in 2D

• Apply functions to different dimensions (axes)

Putting things together

```
What could this be? sin(250x)\frac{1}{\sqrt{2\pi\sigma^2}}e^{-\frac{(x-\mu)^2}{2\sigma^2}} import matplotlib.pyplot as plt time = np.linspace(start=0, stop=0.5, num=2000) mu, sigma = 0.25, 0.01 sinewave = np.sin(time * 250 * np.pi) gaussian = (1 / (np.sqrt(2 * np.pi * np.square(sigma))) * np.exp(-(np.square(time - mu) /np.square(2 * sigma)))) plt.plot(time, gaussian * sinewave) plt.show()
```

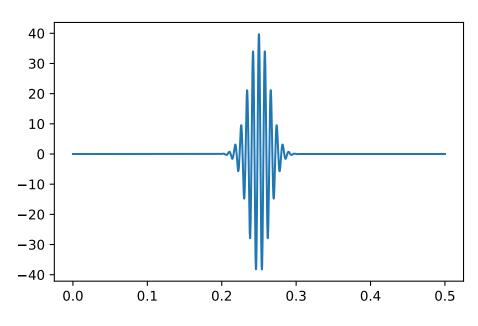


Image for the homework

