

Section 9: Introduction to NumPy and SciPy

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May 24, 2018

Motivation

We have learned all basic data structures...do we need more?

A question

You have an matrix like this:

$$\begin{bmatrix} 1 & 2 & 3 & 4 \\ 4 & 5 & 6 & 7 \\ 7 & 8 & 9 & 10 \end{bmatrix}$$

and you want to sum up numbers by each column. How do you write code for it?

In Python

a solution using native list

```
sums = []
for col_idx in range(len(matrix[0])):
    sum = 0
    for row_idx in range(len(matrix)):
        sum += matrix[row_idx][col_idx]
    sums.append(sum)
print sums
```

In Python

a solution if using numpy arrays

```
print matrix.sum(axis=1)
```

Another comparison

Sum benchmark: summing over a list

```
from numpy import arange
import time

N = 10000000
numpy_array = arange(N)
python_list = range(N)
print "### python list ###"
start = time.time()
sum = 0
for i in python_list:
    sum += i
print "average is: ", float(sum) / N
print "used time: ", time.time() - start
print "### numpy array ###"
start = time.time()
print "average is: ", numpy_array.mean()
print "used time: ", time.time() - start
```

First, import

```
import numpy
```

OR

```
import numpy as np (assuming this from now on)
```

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- ▶ You can use the same indexing:
 - ▶ `a[:2]`
 - ▶ `a[1]`
 - ▶ `a[1:]`
- ▶ arrays can easily be multidimensional: `a = np.array([[1, 2, 3], [4, 5, 6]], float)`

Arrays shapes

```
a.shape == (3, 4)
```

```
[ 1  2  3  4  
 5  6  7  8  
 9 10 11 12]
```

```
a.sum(axis=0)?
```

```
a.sum(axis=1)?
```

Arrays shapes

```
a.shape == (3, 4)
```

1	2	3	4
5	6	7	8
9	10	11	12

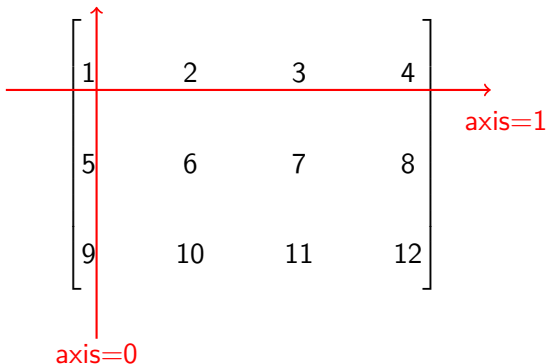
axis=0

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Arrays shapes

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Arrays, reshape

```
a = a.reshape((4, 3))
```

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \\ 10 & 11 & 12 \end{bmatrix}$$

Arrays, reshape

```
a = a.reshape((-1, 6))
```

$$\begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 \\ 7 & 8 & 9 & 10 & 11 & 12 \end{bmatrix}$$

Other ways to create arrays

▶ `a = np.zeros(5)`

```
[0., 0., 0., 0., 0.]
```

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[0., 0., 0., 0., 0.]
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▶ `a = np.arange(0, 10, 2)`

```
[0, 2, 4, 6, 8]
```

Other ways to create arrays

▶ `a = np.zeros(5)`

```
[0., 0., 0., 0., 0.]
```

▶ `a = np.arange(0, 10, 2)`

```
[0, 2, 4, 6, 8]
```

▶ `a = np.full((2, 2), 2)`

```

$$\begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix}$$

```

Array ↔ list conversions

list to array

```
lst = [1, 2, 3]
a = np.asarray(lst)
```

array to list

```
a = np.array([1, 2, 3], int)
lst = a.tolist()
```

Useful operations

- ▶ `sum`, `mean`
- ▶ `np.var`, `np.std`
- ▶ `max`, `min`, `argmax`, `argmin`
- ▶ `zeros_like()`, `ones_like()`
- ▶ `concatenate`

Again, just like `matplotlib`, read the docs!!!

<https://docs.scipy.org/doc/numpy/>

Try it!

Practice

You are given a matrix with each row as a vector. Find the index of the row which has the smallest L_2 norm.

As a reference, for any vector \vec{v} , its L_2 norm is defined as:

$$\|\vec{v}\|_2 = \sqrt{\sum_{k=1}^n v_k^2}$$

example

```
matrix = [[1, 2, 3], [4, 5, 6], [7, 8, 9]]  
return 0
```

A solution with just Python

```
import math

def l2_norm(lst):
    sum = 0
    for i in lst:
        sum += i ** 2
    return math.sqrt(sum)

smallest = None
idx = None
for i in range(len(matrix)):
    l2 = l2_norm(matrix[i])
    if smallest is None or l2 < smallest:
        idx = i
        smallest = l2
print "index: ", i
```

A solution with NumPy

```
import numpy as np
from numpy.linalg import norm

matrix = np.asarray(matrix)
all_norms = norm(matrix, axis=1)
print "index: ", all_norms.argmax()
```

Final word on NumPy: Vectorization

Do not waste NumPy's awesome performance by writing for loops on them!

for loop

```
a = np.arange(10000).reshape((-1, 2))
# square entries
for i in range(len(a)):
    for j in range(len(a[i])):
        a[i][j] = a[i][j] ** 2
```

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```

vectorization code

```
a = np.arange(10000).reshape((-1, 2))
# square entries
a = a * a
```

Now let's switch to SciPy!

```
scipy.cluster
```

```
scipy.constants
```

```
scipy.fftpack
```

```
...
```

```
scipy.signal
```

```
scipy.stats
```

Vector quantization / Kmeans

Physics/Math constants

Fourier Transform

```
...
```

Signal Processing

Statistics

SciPy is built on NumPy

- ▶ You need to know how to deal with NumPy arrays to be comfortable with SciPy functions.
- ▶ Depending on your need, you can almost find anything in it!
- ▶ Commonly used by me: `stats`, `optimize`, `signal`

Optimization: Convex, Non-Convex, ...

optimize module deals with Lagrange multipliers for you!

A convex function

$$\min_x \frac{1}{2}x^2 \text{ s.t. } x \geq -10$$

In SciPy: define objective function, and the constraints

```
def objective(x):  
    return 0.5 * (x ** 2)  
def constraint(x):  
    # unlike definition (<=0), scipy constraints are  
    return x + 10
```

Optimization, cont'd

```
x0 = 0
cons = {'type': 'ineq', 'fun': constraint}
# minimize
minimize(objective, x0, method="SLSQP",
         constraints=cons)
```

Would still work on non-convex constraints such as $\|\vec{v}\|_2 = 0$

Statistics: Student T Test

Hypothesis testing: p values

```
from scipy.stats import ttest_ind
import numpy as np

# two independent random variables
X = np.random.rand(10, 1)
Y = np.random.rand(10, 1)
# T test (two tailed p value)
t, p = ttest_ind(X, Y)
```