

# 資料科學的工具使用（1）

參考資料:Stanford CS231<http://cs231n.github.io/python-numpy-tutorial/>

# 複習一下quicksort

```
def quicksort(arr):  
  
    # recursion都要有一個終點  
    if len(arr) <= 1:  
        return arr  
  
    # 抓中間  
    pivot = arr[len(arr) // 2]  
  
    # 把list依照大小分成三部分  
    left = [ x for x in arr if x > pivot]  
    middle = [ x for x in arr if x == pivot]  
    right = [x for x in arr if x < pivot]  
  
    # 核心概念：  
    # 然後利用遞迴，不斷的一直分成有序三部分，越分越小，  
    # function call會一直stack up，最後全部都分好後，再不斷的組合起來，  
    return quicksort(left) + middle + quicksort(right)  
  
test_arr = [2,3,45,5,333,22,1,2,3,34,4]  
  
quicksort(test_arr)
```

# 基本操作

```
# 確認python版本  
!python --version
```

```
# assign variable  
x = 3
```

```
print(type(x))  
print(x)
```

```
x += 1
```

```
x *= 6
```

```
# python doesn't have X++ as Javascript
```

# AND, OR, NOT, XOR

```
t = True
f = False

print( t and f)
print( t or f)
print( not t)

# XOR
print( t != f)
```

## Basic string op

```
s = "hello"  
# s.capitalize()  
# s.upper()  
# s.replace('l', 'X')  
  
list(range(5))
```

# numpy baisc

```
import numpy as np

list(range(5)) # [0, 1, 2, 3, 4]

a = np.array([1,2,3])
type(a)

# see row,col pair
a.shape

# access element
a[2]
```

cont.

```
# 2 by 3 matrix  
b = np.array([[1,2,3],  
              [4,5,6]])  
  
# check shape  
b.shape  
  
# access element  
b[1,2]  
  
# create diagonal matrix  
c = np.eye(8)  
c
```



## Subarray from original array

```
# get subarray from ori-array,  
# modify the subarray also change the ori-array
```

```
a = np.array([[1, 2, 3, 4],  
              [5, 6, 7, 8],  
              [9,10,11,12]])
```

```
a
```

```
b = a[0:2, 1:3] # row 0 to 2, w/o 2 and col 1 to 3, w/o 3
```

```
b
```

# Integer array indexing

另一個access elt 的操作方法

```
# Integer array indexing

# create a matrix first
a = np.array([[1,2],
              [3, 4],
              [5, 6]])

# Like, row-col pair (0,0), (1,1), (2,0) to access elt
a[[0,1,2],
   [0,1,0]]

# equivalent to this
np.array([a[0, 0], a[1, 1], a[2, 0]])

# a[[0,1,2],
#    [1,0,1]] # 2,3 6
```

# Integer array indexing example

```
# arrange is another way to create list
# np.arange(4)
# np.arange(1,3)
# np.arange(1,6,2)
# np.arange(1,2,0.1)

a = np.array([[1,2,3],
              [4,5,6],
              [7,8,9],
              [10, 11, 12]])

b = np.array([0,2,0,1])

# np.arange is just another way for creating list
a[np.arange(4),b]

# [0,1,2,3]
# [0,2,0,1]

# you can also change the element
a[np.arange(4),b] += 100
a
```

## Boolean index in matrix

```
# so you can create a true/false matrix with given condition
a = np.array([[1,2],
              [3, 4],
              [5, 6]])

b = a > 2
b
# you can also feed this matrix into a and output
# only the true value list / or rank 1 array

a[b]

# or more succinctly
a[a>2]
```

# dot product and elt-wise multiplication

you can define data type with `dtype=np.float64`

another one is `dtype=np.int64`

```
x = np.array([[1,2],
               [3,4]], dtype=np.float64)

y = np.array([[5,6],
               [7,8]], dtype=np.float64)

v = np.array([9,10])
w = np.array([11, 12])

print(x.dot(y))
print(x * y)
```

## sum up array by row/col

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

print(np.sum(x))

# add up col and output np rank1 array
print(np.sum(x, axis=0))
type(np.sum(x, axis=0))

# add up row
print(np.sum(x, axis=1))
```

# transpose

```
# transpose  
  
x = np.array([[1,2],  
              [3,4]])  
x.T
```

# reshape

we use reshape to match elt-wise and conduct outer product

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

# both way are fine
v1 = np.reshape(v, (3,1)) # v.reshape(3,1)

w = np.array([4,5])
print(v1)
print(w)

# outer product
v1 * w
```



# amazing broadcasting

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

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

```
# broadcasting
```

```
# it will auto help me make two matrices the same rank
```

```
# and compute
```

```
x + v
```

# broadcast example

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

```
w = np.array([4,5])
```

```
# x shape is 2,3
```

```
# x transpose will be 3,2 and we can broadcast against w
```

```
# then we transpose back
```

```
# (x.T + w).T
```

```
# or you can reshape w to 2, 1 and broadcast against x
```

```
# np.reshape(w,(2,1)) + x
```

```
# numpy also do array scalar multiplication via broadcasting
```

```
x * 2
```