#### **COMP7035**

#### Python for Data Analytics and Artificial Intelligence

Numpy, Matplotlib, Seaborn

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#### What Will We Learn This Class?

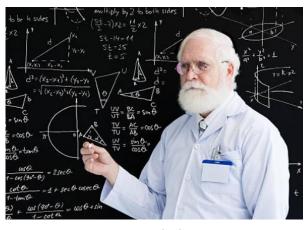
<u>Topic</u>		<u>Hours</u>
I.	Python Fundamentals  A. Program control and logic  B. Data types and structures  C. Function  D. File I/O	12
II.	Numerical Computing and Data Visualization Tools and libraries such as  A. NumPy  B. Matplotlib  C. Seaborn	9
III.	Exploratory Data Analysis (EDA) with Python Tools and libraries such as A. Pandas B. Sweetviz	9
IV.	Artificial Intelligence and Machine Learning with Python Tools and libraries such as A. Keras	9
/2024	B. Scikit-learn	





## Platforms for scientific computing

- Matlab
- Python based platforms
  - Numpy
  - Scipy
  - Matplotlib



Matlab



Python based platforms





# What is Numpy?

- Numpy, Scipy, and Matplotlib provide MATLAB-like functionality in python.
- Numpy Features:
  - Multidimentional arrays (matrices)
  - Fast numerical computations (matrix math)
    - Additional linear algebra, Fourier transform, and random number capabilities
  - High-level math functions



# Why Do We Need NumPy

- Python does numerical computations slowly.
- $1000 \times 1000$  matrix multiply
  - Python triple loop takes > 10 min.
  - Numpy takes ~0.03 seconds
- Arrays are very frequently used in data science, where speed and resources are very important.

$$a = (1,2,3,4,5,6)$$
  $b = (3,4,5,6,7,6)$ 

```
a = [1, 2, 3, 4, 5, 6]
b = [3, 4, 5, 6, 7, 6]
c = []
for t in range(len(a)):
    c.append(a[t]*b[t])
print(c)
```

multiplying each element in a 1-D sequence with the corresponding element in another sequence of the same length

```
a = np.array([1, 2, 3, 4, 5, 6])
b = np.array([3, 4, 5, 6, 7, 6])
c = a*b
print(c)
```



#### How to Use Numpy

• Just import it!

import numpy as np





# Arrays

- Structured lists of numbers
  - Vectors
  - Matrices
  - Images

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

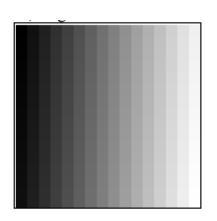
$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix}$$





# Arrays

- Structured lists of numbers
  - Vectors
  - Matrices
  - Images
- What is the relationship between images and arrays?
  - Images can be regarded as a special type of matrics



```
        0
        16
        32
        48
        64
        80
        96
        112
        128
        144
        160
        176
        192
        208
        224
        240

        1
        17
        33
        49
        65
        81
        97
        113
        129
        145
        161
        177
        193
        209
        225
        241

        2
        18
        34
        50
        66
        82
        98
        114
        130
        146
        162
        178
        194
        210
        226
        242

        3
        19
        35
        51
        67
        83
        99
        115
        131
        147
        163
        179
        195
        211
        227
        243

        4
        20
        36
        52
        68
        84
        100
        116
        132
        148
        164
        180
        196
        212
        228
        244

        5
        21
        37
        53
        69
        85
        101
        117
        133
        149
        165
        181
        197
        213
```

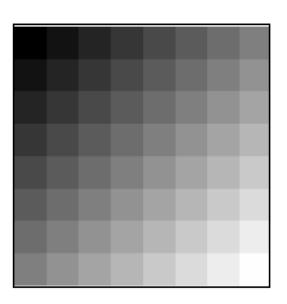






# A Matrix and Its Image

0	50	100	150	200	250	300	350
50	100	150	200	250	300	350	400
100	150	200	250	300	350	400	450
150	200	250	300	350	400	450	500
200	250	300	350	400	450	500	550
250	300	350	400	450	500	550	600
300	350	400	450	500	550	600	650
350	400	450	500	550	600	650	700





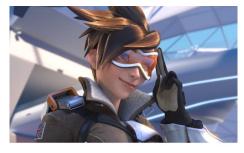


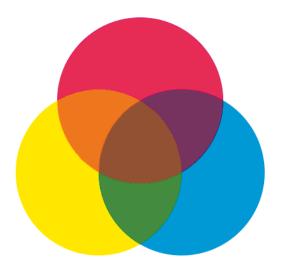
# Image Arrays

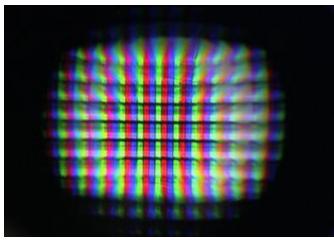
Images are 3D arrays: width, height, and channels Common image formats:

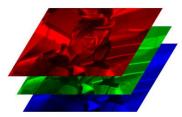
height  $\times$  width  $\times$  RGB (band-interleaved)

height × width (band-sequential)











# Basic Properties of Arrays

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



### How to Create An Array?

- Create a list first.
- Then, please use np.array(list)

```
#E2 python_list = [1,2,3]
np.array(python_list)
```

- Remember how we create a list before?
- We can also create an array using a similar way.

```
#E4

arr = np.array([ 2**i for i in [2,3,9] ])

arr

array([ 4,  8, 512]) (4 8 512)
```



### How to Create An Array?

- Many easy ways can be used to create an array
- We can also create a array using a similar way.
- np.zeros returns a new array of given shape and type, filled with zeros.
- np.ones returns a new array of given shape and type, filled with one.

```
#E6 arr = np.zeros(5)
#E7 arr = np.ones(5)
```





np.ones, np.zeros

```
\#E8 \text{ np.ones}((3, 5), \text{ dtype=np.float32})
```





- np.ones, np.zeros
- np.arange
  - Return evenly spaced values within a given interval
- np.concatenate
- np.zeros\_like, np.ones\_like
- np.random.random

```
#E10
```

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





- np.ones, np.zeros
- np.arange
- You can understand the axis as the axis that will change after the concatenation.
  - Join a sequence of arrays along an existing axis.
- np.zeros\_like, np.ones\_like
- np.random.random

```
#E11
A = np.ones((2,3))
B = np.zeros((3,3))
np.concatenate([A,B],axis=0)
```

```
array([[1., 1., 1.],
[1., 1., 1.],
[0., 0., 0.],
[0., 0., 0.],
[0., 0., 0.]])
```





- np.ones, np.zeros
- np.arange
- np.concatenate
- np.zeros\_like, np.ones\_like
- np.random.random





- np.ones, np.zeros
- np.arange
- np.concatenate
- np.zeros\_like, np.ones\_like
  - Return an array of zeros with the same shape and type as a given array.
  - Return an array of ones with the same shape and type as a given array.
- np.random.random

```
#E12
>>> a = np.ones((2,2,3))
>>> b = np.zeros_like(a)
>>> print(b.shape)
```





#E13

- np.ones, np.zeros
- np.arange
- np.concatenate
- np.astype
- np.zeros\_like, np.ones\_like
- np.random.random
  - Return random floats in the half-open interval [0.0, 1.0). Alias for random\_sample to ease forward-porting to the new random API.

```
>>> np.random.random((10,3))
                                    0.04320502],
array([[ 0.61481644,
                      0.55453657,
         0.08973085,
                      0.25959573,
                                    0.27566721],
                      0.2949532 ,
                                    0.29712833],
         0.84375899,
         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]])
```





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

#### #E14

```
>>> 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

- Total number of elements cannot change.
- Use -1 to infer axis shape, numpy allow us to give one of new shape parameter as -1

#E15

```
a = np.array([1, 2, 3, 4, 5, 6])
print("a:", a)
print("the shape of a:", a.shape)
b = a.reshape(3,2)
print("b:", b)
print("the shape of b:", b.shape)
c = a.reshape(2,-1)
print("c:", c)
print("the shape of c:", c.shape)
```



### Transposition

- np.transpose permutes axes.
- a.T transposes the first two axes. "a is a matrix to be transposed in this place"

```
#E16
a = np.array([[1.,2.],[3.,4.]])
print("a:", a)
b = a.T
print("b:", b)
\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 3 \\ 2 & 4 \end{pmatrix}
```





# Array sorting

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• Sorting is to arrange the elements of an array in hierarchical order either ascending or descending. By default, numpy does sorting in ascending order.

```
#E19

array4 = np.array([1,0,2,-3,6,8,4,7])

array4.sort() [-3 0 1 2 4 6 7 8]

print(array4)
```

```
#E20
array4 = np.array([[10,-7,0, 20],[-5,1,200,40],[30,1,-1,4]])
print("original array \n", array4)
array4.sort()
print("After sorting \n", array4)

10 -7 0 20
sorting -7 0 10 20
-5 1 200 40 -5 1 40 200
```

**-1** 

```
original array
[[ 10 -7 0 20]
[ -5 1 200 40]
[ 30 1 -1 4]]
After sorting
[[ -7 0 10 20]
[ -5 1 40 200]
[ -1 1 4 30]]
```

30





- Arithmetic operations are element-wise
- Logical operator return a bool array
- In place operations modify the array

```
#E17

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

b = np.array([4, 4, 10])

a*b
```

array([ 4, 8, 30])





- Arithmetic operations are element-wise
- Logical operator return a bool array
- In place operations modify the array

```
#E18

a = np.random.random((5, 3))

print("a:", a)

c = a>0.5

print("c:", c)
```

```
a: [[0.1217121 0.97908648 0.8537458 ]
[0.53775343 0.7860607 0.88921186]
[0.853963 0.25478302 0.15270884]
[0.18679235 0.83077973 0.24887868]
[0.29220583 0.43745045 0.91972215]]
c: [[False True True]
[ True True True]
[ True False False]
[False True False]
[False False True]]
```





- Arithmetic operations are element-wise
- Logical operator return a bool array
- In place operations modify the array





- sqrt(), max(), min(), sum(), mean(), std()
  - sqrt() non-negative square-root of an array, element-wise.
  - max() finds the maximum element from an array
  - min() finds the minimum element from an array
  - mean() finds the average of elements of the array
  - std() finds the standard deviation of an array of elements

```
#E22
a = np.array([[1, 4], [9, 16], [25, 36]])
b = np.sqrt(a)
print(b)

[[1. 2.] [3. 4.] [5. 6.]]

#E23
arrayA = np.array([1,0,2,-3,6,8,4,7])
arrayB = np.array([[3,6],[4,2]])
print(arrayA.max()) 8
print(arrayB.max()) 6

#E24
arrayA = np.array([1,0,2,-3,6,8,4,7])
arrayB = np.array([[3,6],[4,2]])
print(arrayA.min()) -3
```

```
#E26
arrayA = np.array([1,0,2,-3,6,8,4,7])
arrayB = np.array([[3,6],[4,2]])
print(arrayA.std()) 3.550968177835448
print(arrayB.std()) 1.479019945774904
```

print(arrayB.min())



• You can find more interesting and important numpy functions from the link below:

https://numpy.org/doc/stable/reference/routines.math.html





# Array Splitting

- numpy.split(ary, indices\_or\_sections, axis=0) splits an array along the specified axis.
  - If indices\_or\_sections is an integer, N, the array will be divided into N equal arrays along axis. If such a split is not possible, an error is raised.

```
array4 = np.array([[10,-7,0,20],[-5,1,200,40],[30,1,-1,4],[1,2,1,0],[0,1,0,2],[0,1,0,2]])
first, second = np.split(array4, 2,axis=0)
print("array4:\n")
print(array4)
print("first:\n", first)
print("second:\n", second)
array4:
[[ 10 -7 0 20]
  -5 1 200 40]
     1 -1 4]
[[ 10 -7 0 20]
[ -5 1 200 40]
[ 30 1 -1 4]]
second:
[[1 2 1 0]
 [0 1 0 2]
 [0 1 0 2]]
```





# Array Indexing

- How to access an element in an array
  - The position for the first element in an array is 0 not 1
  - The position for the last element in an array is -1 or length-1

```
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
arr[5], arr[-3]
(5, 7)
(9, 9)

arr[3:7]
array([3, 4, 5, 6])

array([2, 3, 4, 5, 6, 7, 8, 9])

arr[0:-2]
array([0, 1, 2, 3, 4, 5, 6, 7])
array([0, 2, 4])
```





## Numpy for Image

You need PIL to read image



```
from PIL import Image
import numpy as np

im = np.array(Image.open('data/src/lena.jpg'))
print(type(im))
# <class 'numpy.ndarray'>
print(im.dtype)
# uint8
print(im.shape)
# (225, 400, 3)
```