



14 – NumPy

COMP125 Programming with Python

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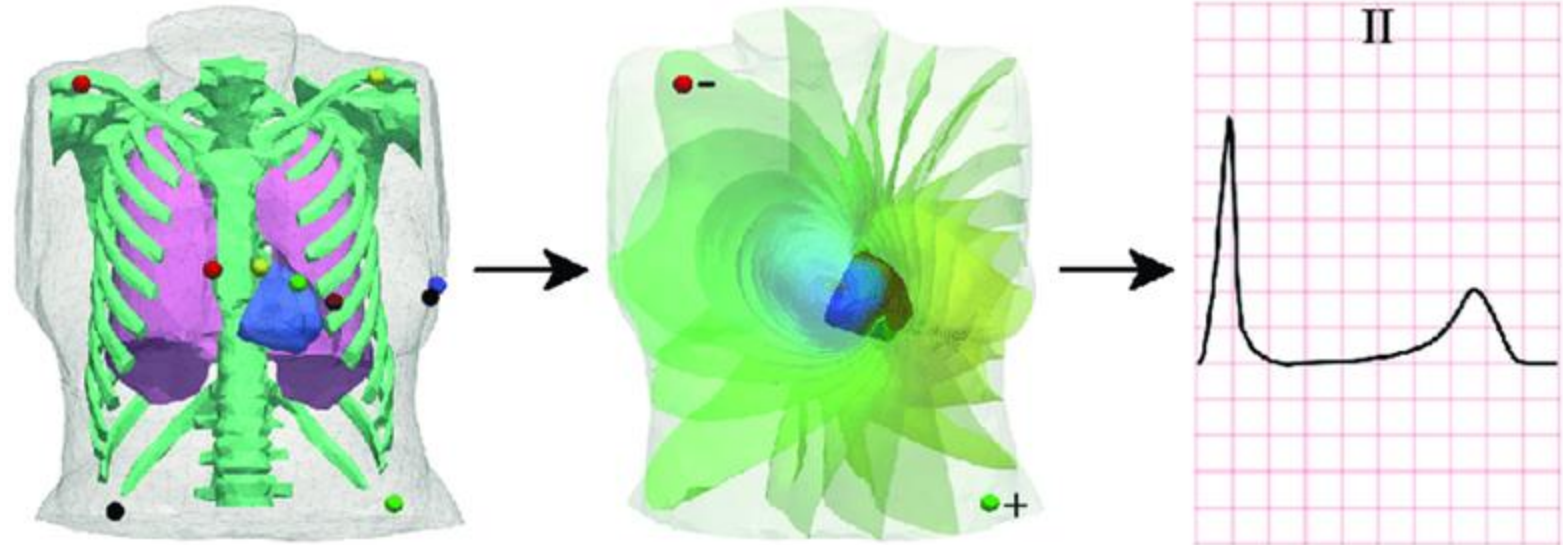
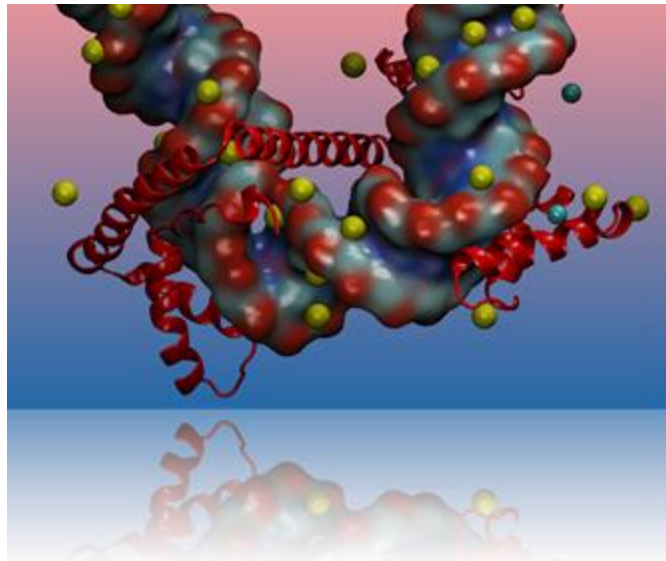
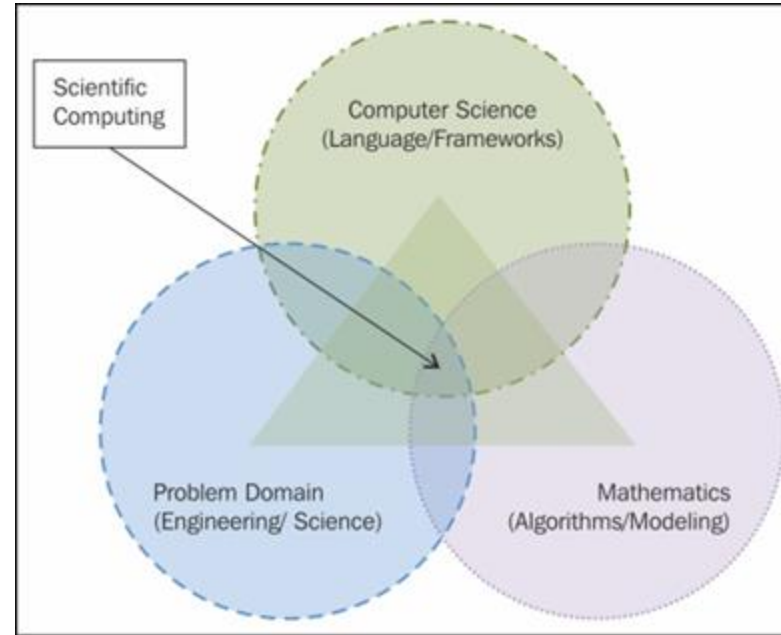
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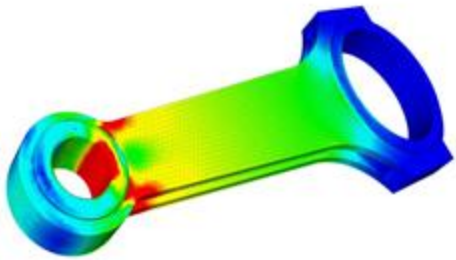
Scientific Computing

- Scientific Computing is the collection of tools, techniques, and theories required to solve mathematical models of problems in Science and Engineering on a computer.

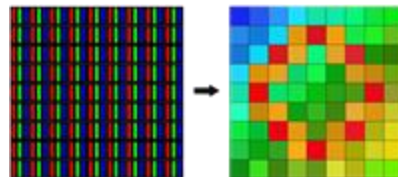


Matrix Operations

- Linear Algebra, Probability and Statistics, and Calculus show ubiquitously in scientific and engineering problems
- Require computers for solving big problems
- Matrix operations are an important subset



Finite
Element/Volume
Methods



#642b4e	#7b4360	#936073
R: 100	R: 123	R: 147
G: 43	G: 67	G: 96
B: 78	B: 96	B: 115

100	123	147
122	161	200
148	202	246

43	67	96
67	114	156
95	155	208

78	96	115
96	122	143
113	145	172

Representing and working on Images

System of linear equations



$$\begin{aligned}
 \text{Joint 2 : } & \begin{cases} f_2 = f_6 \\ f_3 = 10 \end{cases} \\
 \text{Joint 3 : } & \begin{cases} \alpha f_1 = f_4 + \alpha f_5 \\ \alpha f_1 + f_3 + \alpha f_5 = 0 \end{cases} \\
 \text{Joint 4 : } & \begin{cases} f_4 = f_8 \\ f_7 = 0 \end{cases} \\
 \text{Joint 5 : } & \begin{cases} \alpha f_5 + f_6 = \alpha f_9 + f_{10} \\ \alpha f_5 + f_7 + \alpha f_9 = 15 \end{cases} \\
 \text{Joint 6 : } & \begin{cases} f_{10} = f_{13} \\ f_{11} = 20 \end{cases} \\
 \text{Joint 7 : } & \begin{cases} f_8 + \alpha f_9 = \alpha f_{12} \\ \alpha f_9 + f_{11} + \alpha f_{12} = 0 \end{cases} \\
 \text{Joint 8 : } & \begin{cases} f_{13} + \alpha f_{12} = 0 \end{cases}
 \end{aligned}$$

Matrix Operations

- So far, we have seen a few simple examples (Remember Lab09) with
 - list of lists (dense matrices)
 - dictionaries (sparse matrices)
- However, these are inefficient – why?
 - Too general, causing overheads
 - All the axes are not equal (remember getting a column from a list of lists?)
 - Gets worse when generalizing to Tensors
 - Looping over items, difficult to parallelize
- Need a specialized implementation - NumPy



The fundamental package for scientific computing with Python

GET STARTED

NumPy v1.19.0 First Python 3 only release - Cython interface to numpy.random complete

POWERFUL N-DIMENSIONAL ARRAYS

Fast and versatile, the NumPy vectorization, indexing, and broadcasting concepts are the de-facto standards of array computing today.

NUMERICAL COMPUTING TOOLS

NumPy offers comprehensive mathematical functions, random number generators, linear algebra routines, Fourier transforms, and more.

INTEROPERABLE

NumPy supports a wide range of hardware and computing platforms, and plays well with distributed, GPU, and sparse array libraries.

PERFORMANT

The core of NumPy is well-optimized C code. Enjoy the flexibility of Python with the speed of compiled code.

EASY TO USE

NumPy's high level syntax makes it accessible and productive for programmers from any background or experience level.

OPEN SOURCE

Distributed under a liberal [BSD license](#), NumPy is developed and maintained [publicly on GitHub](#) by a vibrant, responsive, and diverse [community](#).

https://numpy.org/devdocs/user/absolute_beginners.html

NumPy (Numerical Python)

- Implemented in C (means fast)
- NumPy `ndarrays`: Efficient storage of a **single type of data**
- These arrays can be of any dimension (e.g., vector 1, matrix 2, tensor n)
- Efficient mathematical and logical operations on arrays
- Efficient linear algebra operations on arrays
- Random number (array) generation
- Foundation of the Python scientific computation stack

NumPy Module

- Typically imported as

```
import numpy as np
```

- Its main data structure, ndarray, is created as:

```
x = np.array(array_like)
```

```
import numpy as np

# 1D ndarray
x1D = np.array((3, 4))

# 2D ndarray
x2D = np.array([[3.3, 4.9], [-1, 3.2], [7.1, -5.7]])

print('Contents of 1D ndarray: ')
print(x1D)

print('Contents of 2D ndarray: ')
print(x2D)
```

```
In [1]: runfile('/Users/c
Contents of 1D ndarray:
[3 4]
Contents of 2D ndarray:
[[ 3.3  4.9]
 [-1.   3.2]
 [ 7.1 -5.7]]
```



```
import numpy as np
L1 = [[1, 2, 3], [4, 5]]
A1 = np.array(L1)
```

```
In [1]: runfile('/Users/cigdem/Desktop/comp125/py Files/untitled3.py', wdir='/
/Users/cigdem/Desktop/comp125/py Files')
/Users/cigdem/Desktop/comp125/py Files/untitled3.py:3: VisibleDeprecationWarning:
Creating an ndarray from ragged nested sequences (which is a list-or-tuple of
lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If
you meant to do this, you must specify 'dtype=object' when creating the ndarray
  A1 = np.array(L1)
```

NumPy Arrays – Basic properties

- **ndim** : Number of dimensions
- **shape** : Size of each dimension
- **size** : Total number of elements
- **dtype** : Type of stored data

```
import numpy as np
```

```
x = np.random.randint(10, size = (3, 4, 5))
print("ndim:", x.ndim)
print("shape:", x.shape)
print("size:", x.size)
print("dtype:", x.dtype)
```

```
y = np.array([3.4, 5])
print("ndim:", y.ndim)
print("shape:", y.shape)
print("size:", y.size)
print("dtype:", y.dtype)
```

```
ndim: 3
shape: (3, 4, 5)
size: 60
dtype: int64
```

```
ndim: 1
shape: (2,)
size: 2
dtype: float64
```


NumPy Arrays – Indexing (1D arrays)

```
import numpy as np

L = [10, 20, 30, 40]
for i in range(len(L)):
    print(L[i])

A = np.array(L)
for i in range(len(A)):
    print(A[i])

B = np.array(L)
for i in range(B.shape[0]):
    print(B[i])
```

Nam 	Type	Size	Value
A	Array of int64	(4,)	[10 20 30 40]
B	Array of int64	(4,)	[10 20 30 40]
i	int	1	3
L	list	4	[10, 20, 30, 40]

NumPy Arrays – Indexing (2D arrays)

```
import numpy as np

L = [[11, 22, 33], [44, 55, 66]]
for i in range(len(L)):
    for j in range(len(L[i])):
        print(L[i][j])

A = np.array(L)
for i in range(len(A)):
    for j in range(len(A[i])):
        print(A[i][j])

B = np.array(L)
for i in range(B.shape[0]):
    for j in range(B.shape[1]):
        print(B[i, j])
```

Nan ▲	Type	Size	Value
A	Array of int64	(2, 3)	[[11 22 33] [44 55 66]]
B	Array of int64	(2, 3)	[[11 22 33] [44 55 66]]
i	int	1	1
j	int	1	2
L	list	2	[[11, 22, 33], [44, 55, 66]]

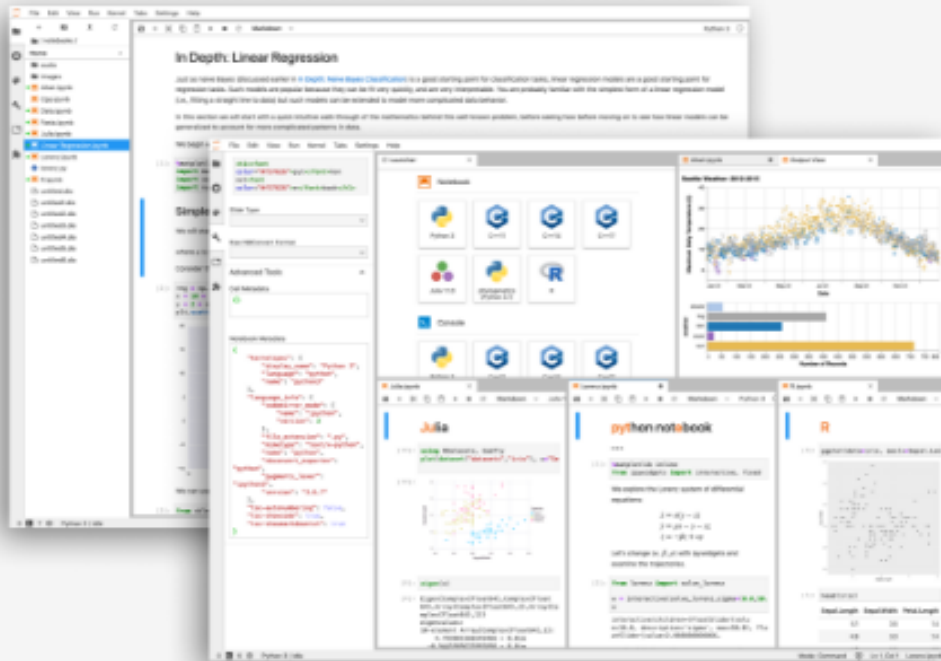
We may also use another way of indexing

JupyterLab

<https://jupyter.org/>



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JupyterLab is a web-based interactive development environment for notebooks, code, and data. Its flexible interface allows users to configure and arrange workflows in data science, scientific computing, computational journalism, and machine learning. A modular design allows for extensions that expand and enrich functionality.

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Applications on

base (root)

Channels

Run a terminal with your current environment from Navigator activated

Launch

Online data analysis tool with more coding assistance by JetBrains. Edit and run your Python notebooks in the cloud and share them with your team.

Launch

IBM Watson Studio Cloud provides you the tools to analyze and visualize data, to cleanse and shape data, to create and train machine learning models. Prepare data and build models, using open source data science tools or visual modeling.

Launch



JupyterLab

3.2.1

An extensible environment for interactive and reproducible computing, based on the Jupyter Notebook and Architecture.

Launch



Notebook

6.4.5

Web-based, interactive computing notebook environment. Edit and run human-readable docs while describing the data analysis.

Launch



Powershell Prompt

0.0.1

Run a Powershell terminal with your current environment from Navigator activated

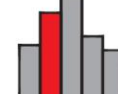
Launch



Qt Console

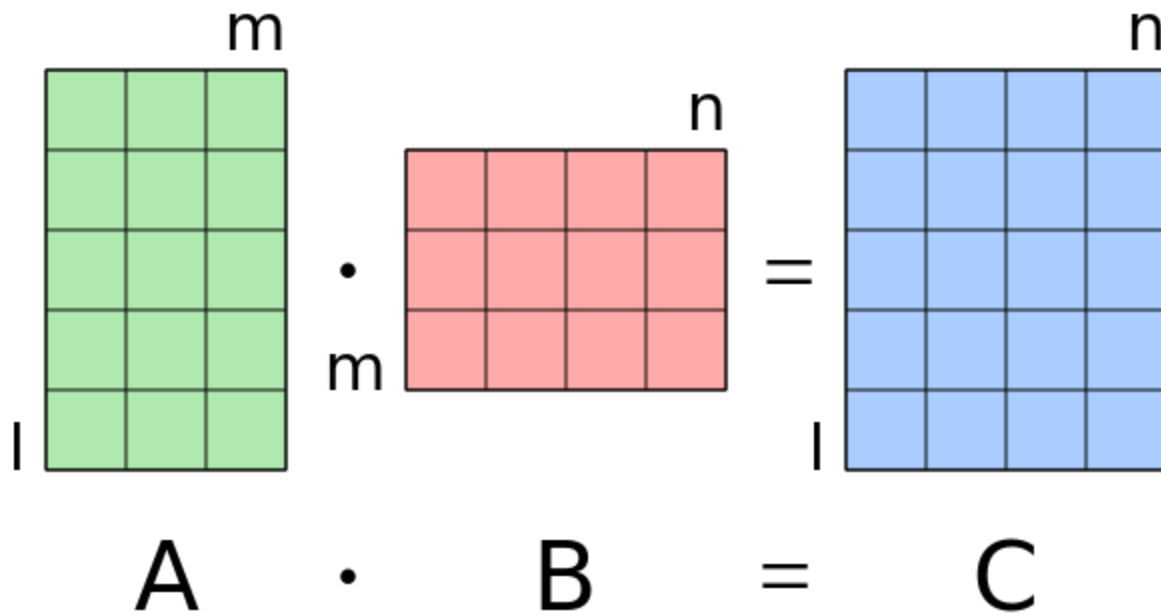


Spyder



Glueviz

Matrix Multiplication



NumPy:

- `np.dot(A, B)`
- `A.dot(B)`
- `A@B`

$$\begin{pmatrix} 0 & 1 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 4 & 5 \\ 6 & 7 \end{pmatrix} = \begin{pmatrix} 6 & 7 \\ 26 & 31 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 4 & 5 \\ 6 & 7 \end{pmatrix} = \begin{pmatrix} 6 & 7 \\ 26 & 31 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 4 & 5 \\ 6 & 7 \end{pmatrix} = \begin{pmatrix} 6 & 7 \\ 26 & 31 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 4 & 5 \\ 6 & 7 \end{pmatrix} = \begin{pmatrix} 6 & 7 \\ 26 & 31 \end{pmatrix}$$