# ENGINEERING PROBLEM SOLVING IN PYTHON

EMCH 561: Machine Learning for Mechanical Engineers





## **Python**

Python is an **interpreted** (mostly), **high-level**, **general-purpose** programming language with an **object-oriented** approach.

- Interpreted language: is a type of programming language for which its instructions execute directly and freely, without previously compiling a program into machine-language instructions.
- **High-level:** a programming language with strong abstraction from the details of the computer.
- **General-purpose:** is a programming language designed to be used for writing software in the widest variety of application domains (a general-purpose language).
- **Object-oriented:** is a programming paradigm based on the concept of "objects", which can contain data, in the form of fields (often known as attributes or properties), and code, in the form of procedures (often known as methods).





## **Python**

Python's design philosophy emphasizes **code readability** with its notable use of significant **whitespace**. Python is **dynamically typed** and **garbage-collected**.

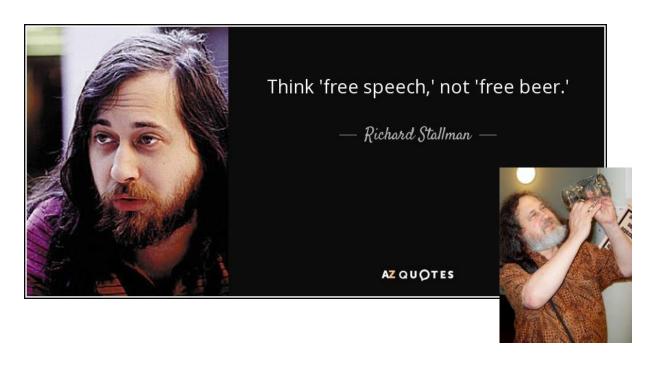
- Code readability: refers to the ease with which a human reader can comprehend the purpose, control flow, and operation of source code. It affects the aspects of quality, including portability, usability and most importantly maintainability.
- Whitespace: refers to the usage of whitespace in typing code. In Python, the information about blocks are expressed by their indentation.
- **Dynamically Typed:** refers to languages execute many common programming behaviors at runtime that static programming languages perform during compilation. For us, this means defining variable types.
- **Garbage-Collected:** is a form of automatic memory management. The garbage collector attempts to reclaim garbage, or memory occupied by objects that are no longer in use by the program.



## Python is Free as in beer and Free as in speech

Free beer (never)



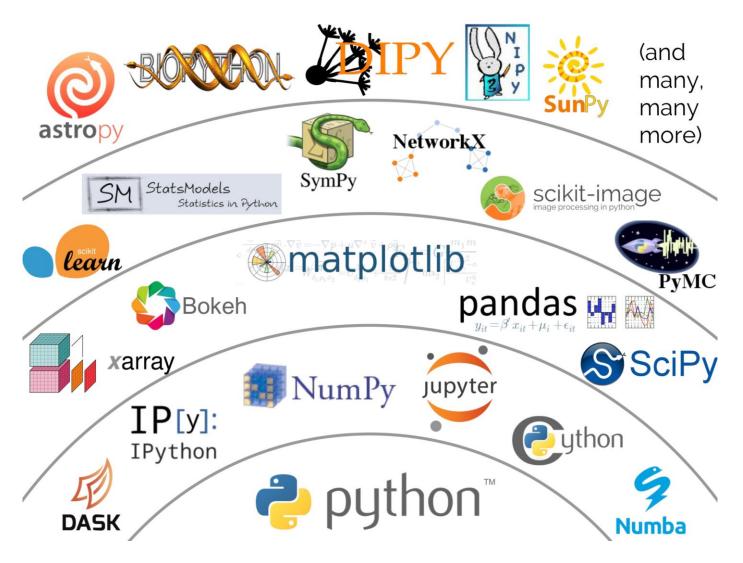


Free speech (First Amendment to the United States Constitution; December 15, 1791)





#### Python is a language, we use an ecosystem



In this class we will focus on a limited subset of this ecosystem:

- IPython: command shell for interactive computing in Python.
- **Spyder:** cross-platform integrated development environment (IDE) for scientific programming in Python.
- Numby: library adding support for multi-dimensional arrays.
- SciPy: library used for scientific computing and technical computing.
- matplotlib: API focused on plotting
- scikit-learn: library for machine learning.
- TensorFlow: symbolic math library, used for neural networks.



### Python Distributions

A distribution is a group of packages that are assembled and more-or-less work well together. Some of these are free (beer/speech) while for others there are required service packs for their use. In this class, we will use Anaconda from Anaconda, Inc. Anaconda is a free and open-source distribution for scientific computing.











#### Why not just use MATLAB?

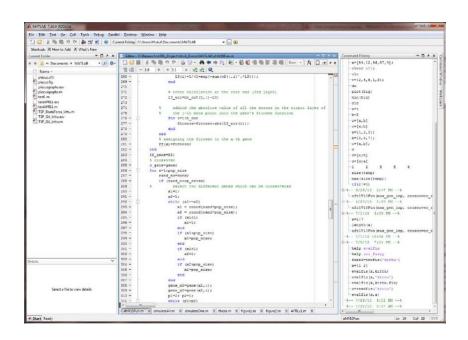
#### Pros:

- Works out of the box.
- Code is well managed and packages don't conflict.
- Widely used in engineering and computing.
- Well formulated for linear algebra with simple commands
- Fast implementation of code for complex problem solving
- Not at the bleeding edge of machine learning

#### Cons:

- Not free as in speech, not free as in beer
- Each module must be paid for separately
- Not at the bleeding edge of machine learning
- Difficult to keep license updated on multiple machines
- Lets you write poorly structured code.
- Code is slower than Python (personal experience)







#### MATLAB does not use whitespace

#### Python

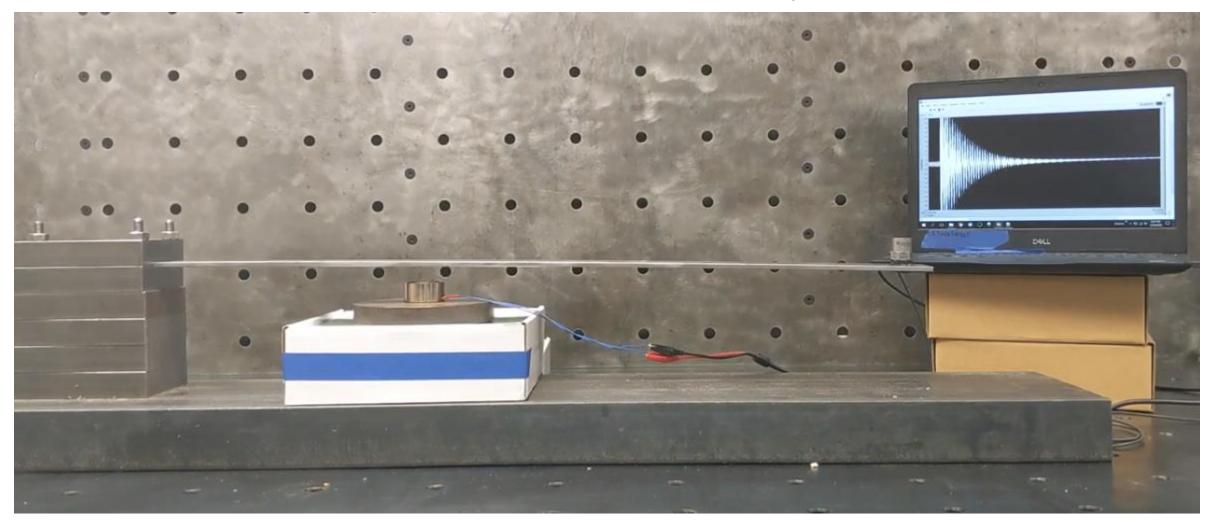
```
whitespace.py """
from random import randrange
def numberizer():
 →# Generate a random number from 1 to 10.
return randrange(1, 11)
number = numberizer()
if number > 5:
print("This number is big!")
class RandomNumberHolder(object):
← → # Create and hold 20 random numbers using numberizer
 →def init (self):
      →self.numbers = [numberizer(x) for x in range(20)]
random_numbers = RandomNumberHolder()
```

#### **MATLAB**

```
% counters for recording results
28
29 -
       dd=1:
        hh=1:
31
      for history = history lengths
33
       % independent variable is the acceleration data
34
       x = data.acceleration data';
36
37
       % dependent variable is the pin location
38 -
       y = data.measured pin location';
39
40
       % delete nan values from pin position
41 -
       nanidx=find(isnan(y));
        for i=nanidx
43 -
       legitidx=find(~isnan(y(1:i)));
44 -
       last legitidx=legitidx(end);
       y(i)=y(last_legitidx);
45 -
46 -
        end
        end
```



## Example #1 – Plot beam vibration data in Python

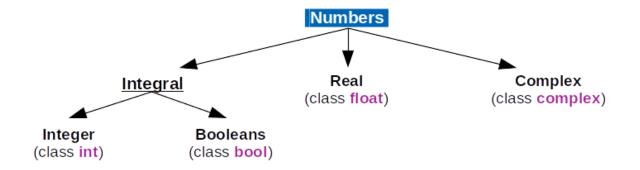


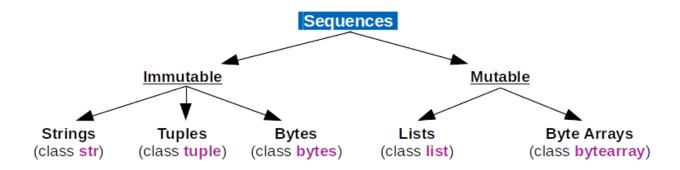


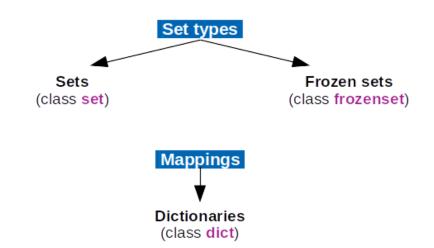
## Python

#### Key types in Python are:

- Numbers:
  - Integer
  - Booleans
  - Real
- Sequences:
  - Strings " or ""
  - Lists []
  - Tuples ()
- Mappings
  - Dictionaries { }







## Example #2 – Pure Python

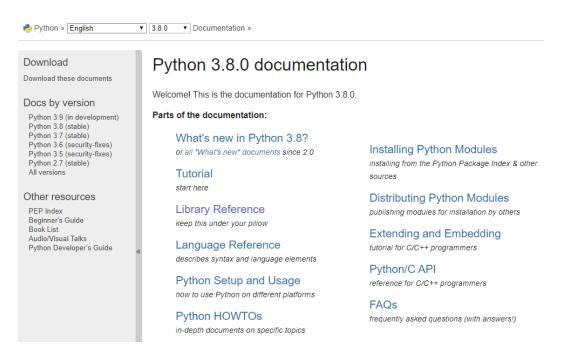




## Additional Python Resources

Python documentation

https://docs.python.org/3.8/

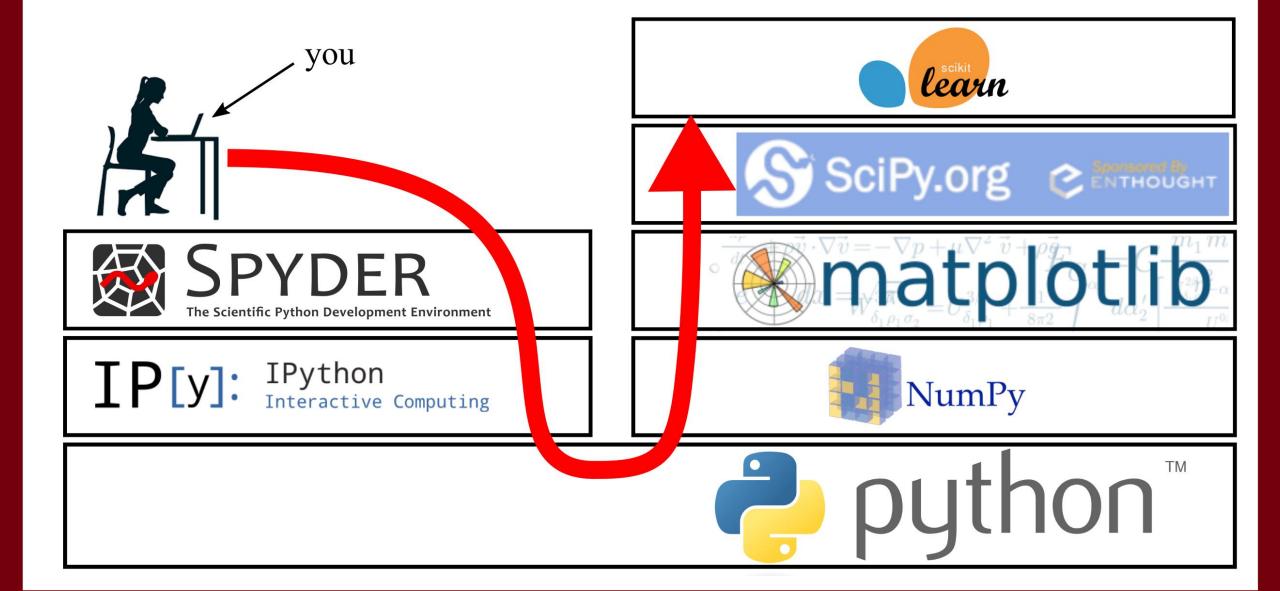


YouTube: "Python Programming" by Derek Banas <a href="https://www.youtube.com/watch?v=N4mEzFDjqtA">https://www.youtube.com/watch?v=N4mEzFDjqtA</a>



And many many more.....

#### Our Ministack

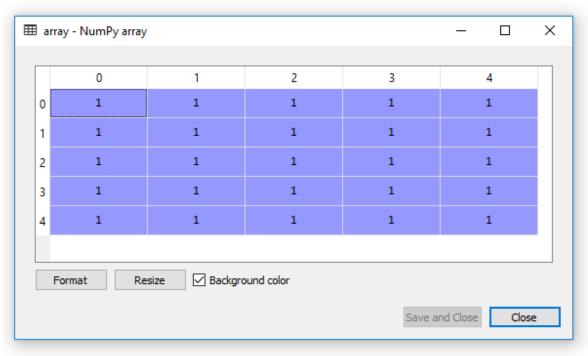


## NumPy (Numerical Python)

- Python does not support arrays or matrices and its mathematical support is limited and slow.
- NumPy was created to add support for arrays and matrices to Python.
- Numpy adds support for high-level mathematical functions to operate on these arrays and matrices.
- Python documentation can be found at <u>https://numpy.org/doc/1.17/reference/index.html</u>
- Numpy is limited in scope and is only designed to add numerical computing to phyton.
- NumPy underpins nearly every scientific computing platform created for Python.



```
39 # import numpy as np
40 import numpy as np
41
42 # build an array
43 array = np.ones((5,5))
```

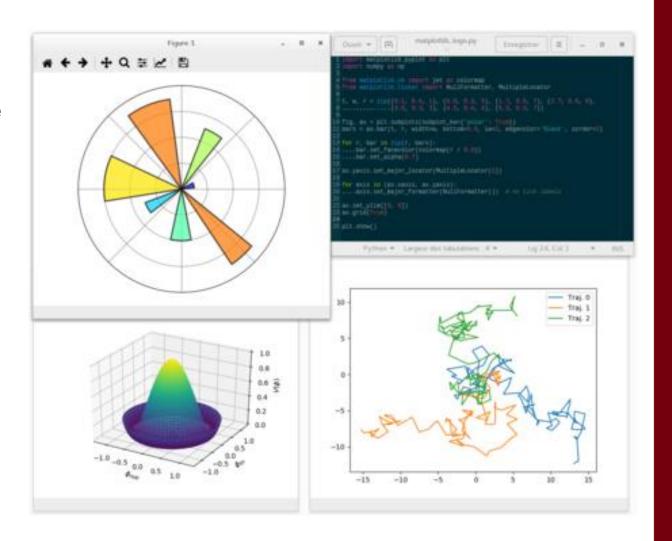




#### Mapplotlib

- Matplotlib is a plotting library for Python.
- At one point it was formed by combining multiple liberies (pyplot and pylab). We will use pyplot as pylabs use is discouraged.
- Pyplot is a Matplotlib module which provides a MATLAB-like interface.
- matplotlib documentation can be found at https://matplotlib.org/3.1.1/api/index.html
- Pyplot documentation can be found at <a href="https://matplotlib.org/3.1.1/api/">https://matplotlib.org/3.1.1/api/</a> as gen/matplotlib.pyplot





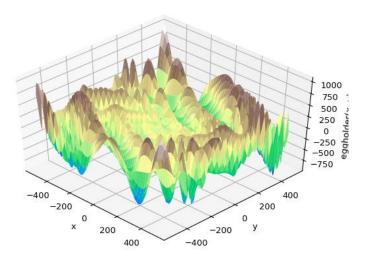


## SciPy (not ScientificPython but close enough)

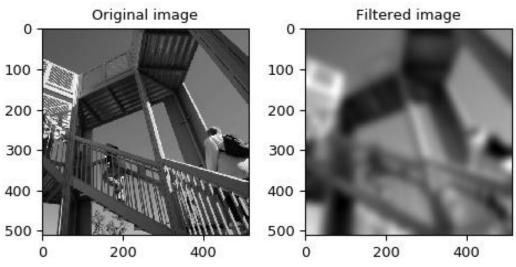
- SciPy is a Python library used for scientific computing and technical computing.
- SciPy builds on the NumPy array class and the capabilities of matplotlib.
- SciPy contains modules for optimization, linear algebra, integration, interpolation, special functions, FFT, signal and image processing, ODE solvers and other tasks common in science and engineering.
- SciPy and NumPy contain some duplicate capabilities (e.g., FFT). A common way to look as this is that the NumPy version is more bare-bones than the equivalent SciPy version.
- SciPy documentation can be found at <u>https://docs.scipy.org/doc/scipy/reference/</u>



#### **Optimization (scipy.optimize)**



#### Signal Processing (scipy.signal)

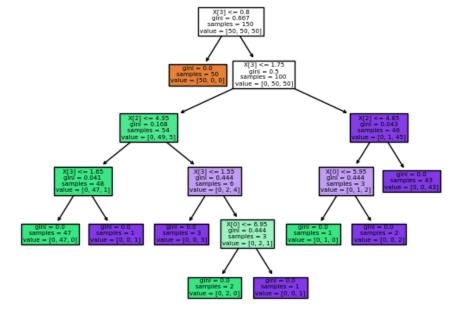


### scikit-learn (also called sklearn)

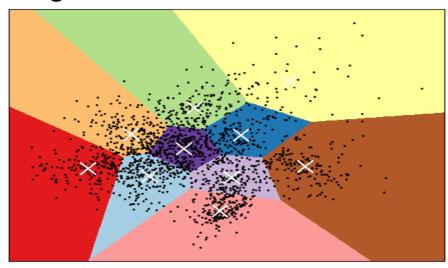
- Sklearn is a machine learning library for Python.
- It features various classification, regression and clustering algorithms including support vector machines, random forests, gradient boosting, k-means and DBSCAN.
- Sklearn is designed to interoperate with the Python numerical and scientific libraries NumPy and SciPy.
- Sklearn documentation can be found at
- https://scikit-learn.org/stable/user\_guide.html



#### **Decision Trees**



#### Clustering



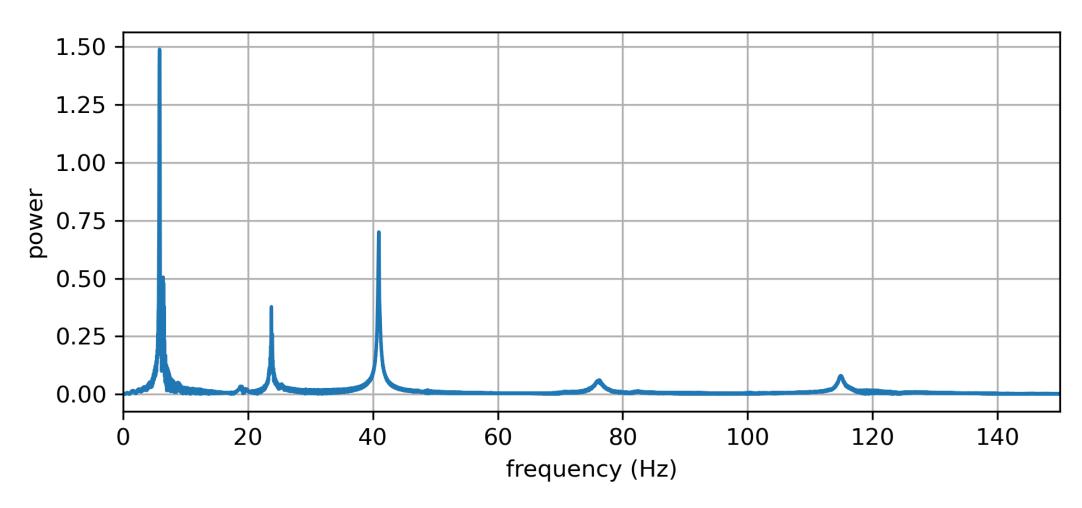
#### **Conclusions:**

- Python is great, but not useful for scientific computing.
- Scientific modules have been developed for python to expand its usefulness into scientific computing. The major packages are:
  - NumPy
  - Matplotlib
    - Pyplot
  - SciPy
  - Sklearn
- We code in a Integrated Development Environment (IDE) of which there are lots, we used Spyder in our examples.

```
@author: austin
import IPython as IP
IP.get_ipython().magic('reset -sf')
import numpy as np
import scipy as sp
import matplotlib as mpl
import matplotlib.pyplot as plt
plt.close('all')
#%% Load and plot data
D = np.loadtxt('vibration_data/Vibration_measurement.csv',delimiter=',')
tt = D[:,0]
dd = D[:,1]
plt.figure('beam data',figsize=(3.5,3))
plt.plot(tt,dd,'-',label='data 1')
plt.grid(True)
plt.xlabel('time (s)')
plt.ylabel('voltage (v)')
plt.title('beam data')
plt.xlim([-0.1,1.6])
plt.legend(framealpha=1,loc=8)
plt.tight layout()
plt.savefig('plot.png',dpi=300)
array = np.ones((5,5))
#%% Plot an FFT of the data
```



## Example #3 – Data processing (FFT) in SciPy





## THANKS!

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