

# Introductory Python Tutorial with Resources

Physics 512 Computational Physics

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# Tutorial Overview

- Basic Unix & Github commands
- Brief review of
  - Data Types
  - Assignments
  - Functions
  - Conditionals
  - Loops
  - Useful Numpy features
  - Matplotlib 1D and 2D plotting
- Practice with a few sample coding tasks

Certain slides are courtesy of Orion Lyau and Nicholas Rui.

# Data Types in Python

Type	Example
<code>int</code>	<code>10</code>
<code>float</code>	<code>3.14</code>
<code>str</code>	<code>"Cat", 'any text'</code>
<code>bool</code>	<code>True, False</code> (note first capital letter)

More thorough introduction to data types: <https://realpython.com/python-data-types/>

# Data Types in Python

Type	Example
tuple	<code>(1, 3, 4)</code> <small>(note immutable)</small>
list	<code>['Cat', 'Dog', 2]</code>
dict	<code>{'key': 'value', 'coding': 'fun'}</code>
set	<code>{'Apple', 'Banana', 'Orange'}</code>

More thorough introduction to data types: <https://realpython.com/python-data-types/>

# Assignment

Assignment takes the **expression** on the right hand side of the = and **binds** it to the **name** on the left hand side of the =. The = is the assignment **operator**.

```
pi = 3.15159
```

```
tau = 2 * pi
```

This is the simplest means of **abstraction**. Here we can use names to refer to more complicated things.

# Type Conversion

Example: `str` to `int`

```
my_int = 1
```

```
my_string = str(my_int)
```

Example: `int` to `str`

```
new_string = '8'
```

```
new_int = int(new_string)
```

# Anatomy of a Function

Functions are defined in the following format:

```
def <name>(<parameters>):  
    return <expression>
```

It is important to indent each line underneath the **def** statement

# Anatomy of a Function

Example:

```
def hypotenuse(a, b):  
    return (a ** 2 + b ** 2) ** 0.5
```

Name

Parameters

Expression to return

Usage:

```
hypotenuse(3, 4)
```

Arguments



# Scope

When you're in the body (indented region) of a function, the variables you create only exist inside that function (i.e. within the **scope** of the function).

```
def times_five(x):  
    result = x * 5  
    return result
```

```
def times_ten(x):  
    result = x * 10  
    return result
```

The **result** in **times\_five** and **times\_ten** are independent of each other, and each **result** is not accessible outside their respective functions.

# Comparison Operators

Comparisons evaluate to **True** or **False** and include the following operators:

`==` equal to

`!=` not equal to

`>` greater than

`<` less than

`>=` greater than or equal to

`<=` less than or equal to

Examples:

`1 == 1` → True

`1 != 1` → False

`5 > 3` → True

Operator

Operand

# Boolean Operators

Boolean operators (sometimes also called **logical operators**) evaluate to **True** or **False** and include the following operators:

**and**  
**or**  
**not**

Examples:

True and False	→	False
True or False	→	True
not False	→	True

# Conditional Statements

Example:

Stuff inside conditional statements must be indented.

**if** <expression>:

    <stuff>

Starts at the top, and the only the **<stuff>** under the first true expression is run.

**elif** <expression>:

    <stuff>

Must start with **if** clause, followed by zero or more **elif** clauses, and zero or one **else** clause.

**else**:

    <stuff>

If an **else** clause is present, the stuff there will only run if all **if** and **elif** expressions evaluate to false.

# while loop

A **while** loop repeats for as long as a given expression is true.

Example:

```
while <expression>:  
    <stuff>
```

1. Evaluate <expression>, and if <expression> is true, go to step 2
2. Run <stuff>, then go back to step 1

You need to make sure that <expression> will eventually evaluate to false, otherwise the loop will repeat indefinitely. If this happens, it is said to be an **infinite loop**.

# for loop

A **for** loop repeats as many times as there are entries in the thing you're iterating over.

Example:

```
for <name> in <iterable>:  
    <stuff>
```

At each iteration, you will be able to access the current element of **<iterable>** through the loop variable **<name>**.

Things you might often iterate over include lists, tuples, or ranges of numbers. The latter can be represented by **range(n)** where **n** is the end of the range.

# break and continue

In a **for** or **while** loop, the **break** and **continue** statements let you either terminate a loop (**break**) or continue to the next loop iteration (**continue**).

You can also use the **return** statement anywhere within a function (including within a loop) to immediately exit the function and return a value, without needing to execute anything further.

# import numpy as np

Numpy is the core scientific computing package in Python!

- Numpy arrays (`np.array`) are very similar to Python lists, but with extra features!

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

- `a+1` # Prints ``array([2,3,4])``
- `a*2` # Prints ``array([2,4,6])``
- `a.shape` # Prints `(3,)`
- `a[0]` # Prints ``1``

NumPy has many great functionalities to explore! See the [documentation](#)



# Ex: Array creation

There are a number of ways to generate numpy arrays:

- `a = np.array([[1,2],[3,4]])` # From a (nested) list
- `b = np.zeros((2,2))` # Generates a 2x2 array of 0s
- `c = np.ones((2,2))` # Generates a 2x2 array of 1s
- `d = np.full((2,2),5)` # Generates a 2x2 array of 5s
- `e = np.eye(2)` # Generates a 2x2 identity matrix

... And more! You can also do element-wise math between arrays:

- `a+c` # Prints ``array([[2., 3.], [4., 5.]])``
- `a*b` # Prints ``array([[0., 0.], [0., 0.]])``

... etc!

# 1D/2D Linear/Logarithmically Spaced Arrays

Often we want to generate a series of equally spaced numbers in numpy to feed into a function  $f(x)$  or for other uses. This is simple enough:

- `x = np.linspace(0,1,100)` # 100 elements spaced btw 0 & 1
- # NOTE: `np.logspace` also exists, and does what you think!
- `y = f(x)` # where  $f$  is your desired function

For functions of two variables,  $f(x,y)$ , `np.meshgrid` is also super useful!

- `x = np.linspace(0,1,3); y = np.linspace(0, 2, 2)`
- `X, Y = np.meshgrid(x,y)`
- `Z = f(X, Y)` # returns function evaluated on the grid

# Matrix Operations

NumPy can easily handle pretty much any matrix operation

- `a = np.array([[1,2],[3,4]]); b = np.array([[5,6],[7,8]])`
- `a @ b` # this is the matrix multiplication of a & b
- # NOTE: recall `a * b` is element-wise multiplication!
- `a_inv = np.linalg.inv(a)` # matrix inversion
- `e_val, e_vec = np.linalg.eig(a)` # eigenvalues/vectors

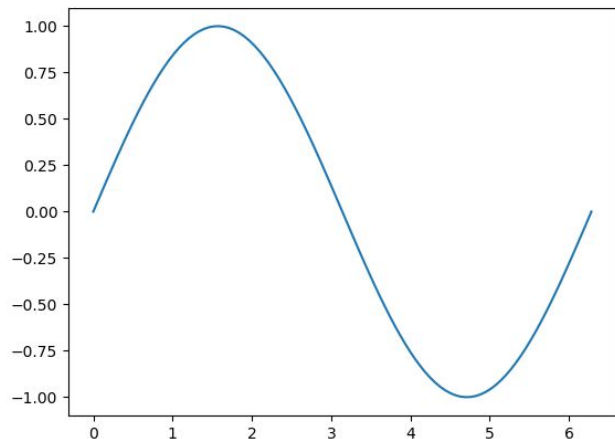
`np.linalg` has a lot of interesting functionalities, and if you want to learn more you can explore what the library has to offer [here](#)

# import matplotlib.pyplot as plt

[Matplotlib](#) is the package pretty much everyone uses for plotting data in python. In particular, `matplotlib.pyplot` is the function you'll be using most of the time, and pretty much everyone import it as `plt`!

- `x = np.linspace(0,2*np.pi,100); y = np.sin(x)`
- `plt.plot(x,y); plt.show()`

This will very easily produce a simple plot:



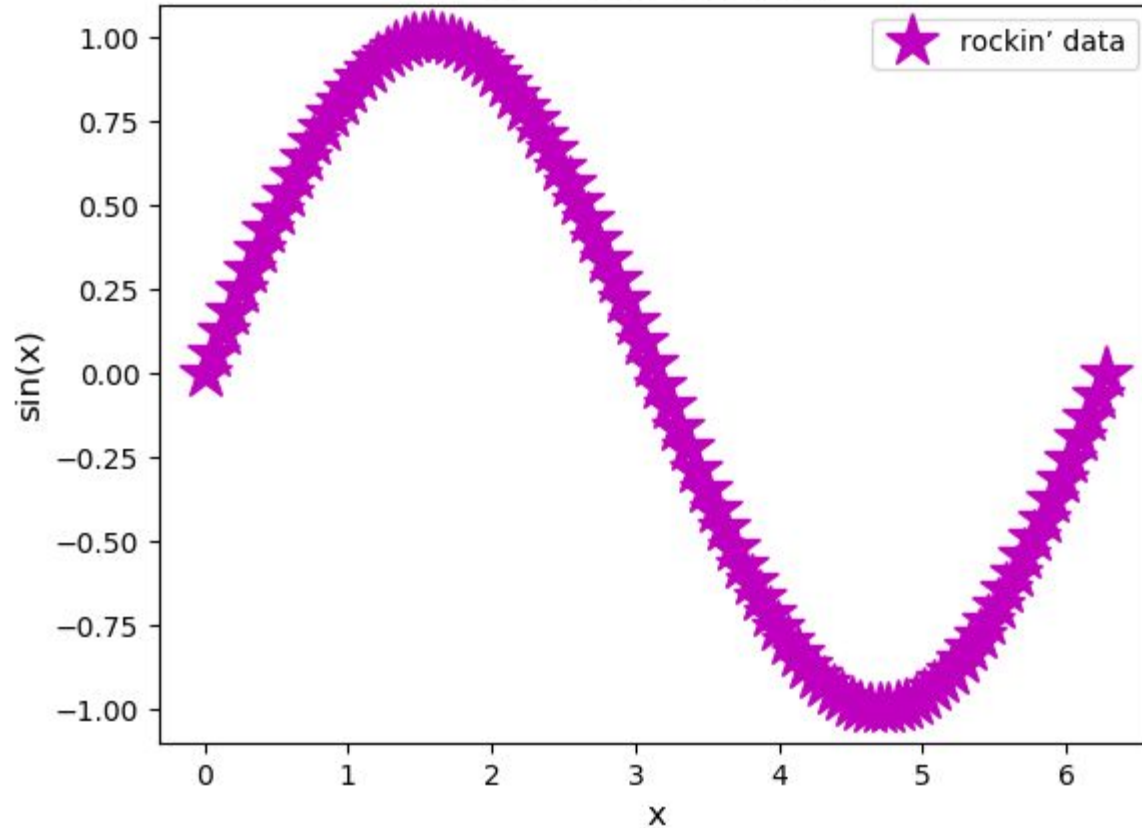
# AESTHETIC

“But Marcus!”, you object, “your axes have no labels, and the plot is so bland!”  
Well, this is all entirely correct, so let’s spice up our plot a bit

- `fmt = 'm*'; size=20; label="rockin' data"`
- `plt.plot(x,y,fmt=fmt,markersize=size,label=label)`
- `plt.legend(); plt.xlabel("x",fontsize="large")`
- `plt.ylabel("sin(x)", fontsize="large")`
- `plt.title("much better ;)", fontsize="xx-large")`

Which gives us...

much better ;)



Obviously your day-to-day plots don't have to be as extravagant as... *this*, but it's worthwhile to play around with the formatting options in `plt.plot` to find the best way to represent your data!

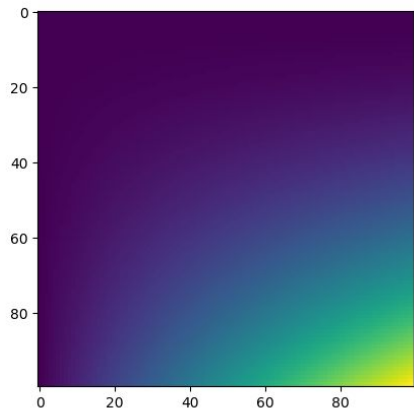
# 2D Plots

Now say we want to plot a function of two variables! Matplotlib actually has a number of ways to do this, each with its advantages and disadvantages.

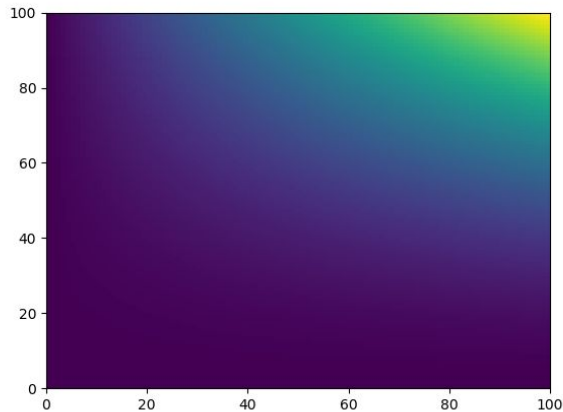
- `x = np.linspace(0,1,50); X, Y = np.meshgrid(x,x)`
- `Z = X * Y**2 # some arbitrary 2D function`
- `plt.imshow(Z);plt.show()`
- `plt.pcolormesh(Z);plt.show()`
- `plt.contourf(Z);plt.show()`

NOTE: There are many [colour maps](#) to choose from and I guarantee at some point in your life you'll spend way too much time experimenting to find juuuust the right one! But more seriously, a good choice of colour map is important.

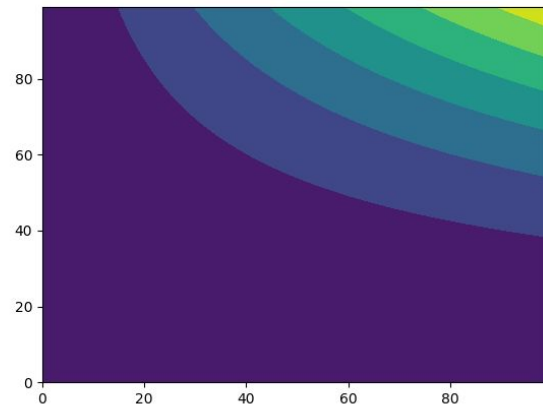
# imshow



# pcolormesh



# contourf



Each of these functions has their advantages and disadvantages, and I recommend doing a bit of reading on each to figure out which is best for your use case.

If you're curious, I find myself using `imshow` the most of all of them, because it's (in my opinion) the simplest.

Also, the colour bars are missing, but a simple `plt.colorbar()` call before showing/saving the figure will remedy this!



That's all!

Any questions before some exercises?

# Resources

Very new to Python? Try reading the first few chapters of this book for a thorough introduction to Python: [Python for Astronomers](#)

[NumPy Basics](#)

[From Python to NumPy](#) (more advanced)

[The Python Data Science Handbook](#)

[Some practice tutorials](#)

[Some examples of animations with matplotlib](#)

[Computer Memory](#) (as discussed in first lecture)