

## Python 03

Elias Wachmann

2024



#### Content

- 1. Variables
- 2. Python files
- 3. Functions
- 4. Fitting Data



## **Variables**



#### Variable Names

#### From PEP-8 Style Guide:

- must start with letter or \_
- can only contain alpha-numeric values (a-z, A-Z, 0-9)
   and \_
- variable names are case sensitive
- do not overwrite built-in functions



#### Naming conventions

#### According to PEP-8 Style Guide:

- CONST\_NAMES are all caps with \_ between words
- function\_names are all lower case with \_ between words
- variables follow the same rules as functions
- ClassNames are CamelCase



# Python files



## Python files and usage

- Python files are called .py
- Python files can be run from the command line
  - python3 file.py
- Python files can be imported into other Python files
  - import numpy
  - or with an alias import numpy as np
- Python files can be run as scripts or imported as modules



## Python files run as scripts

I can simply run this file and it will print the sinus values from y for the indices 5 through 10 (excluding 10).

```
# myscript.py
import numpy as np

x = np.arange(0, 10, 0.1)
y = np.sin(x)
print(y[5:10])
```

What happens if I import this file in another file?



### Python files imported as modules

I can now import myscript.py in another file using the import statement:

```
import myscript
print("This is import.py")
```

What happens if I run this file?



### Python files imported as modules

```
lides] % /bin/python3 /home/etschgi1/Desktop/REPOS/exercises-python/slide
s24/03/examples/import.py
[0.47942554 0.56464247 0.64421769 0.71735609 0.78332691]
This is import.py
```

It not only prints the *This is import.py* line, but also the sinus values from y for the indices 5 through 10 (excluding 10).

These are printed because the import statement runs the file as a script.

How to avoid this?



\_\_name\_\_ == '\_\_main\_\_'

By using the \_\_name\_\_ variable, we can avoid running the file as a script when it is imported as a module.

```
# myscript2.py
import numpy as np

if __name__ == "__main__":
    x = np.arange(0, 10, 0.1)
    y = np.sin(x)
    print(y[5:10])
```

Importing this file in another file will not print the sinus values.



## **Functions**



#### **Functions**

- Functions are defined with def
- Functions can have arguments
- Functions can return values
- Functions can have default arguments
- Functions can have variable number of arguments



#### Some examples

```
def hello():
      print("Hello, world!")
2
3
4
 def arguments(a, b, c):
      print("Got arguments:", a, b, c)
6
7
8
  def return_value():
      return 42
10
```



#### Some examples

```
1 def add(a, b=2):
2    return a + b
3
4
5 add_two_arguments = add(30, 12)
6 add_one_argument = add(40)
```



#### Multi returns

#### Multiple returns from the same funciton

```
def multireturn(a, b):
    sum_ = a + b
    product = a * b
    return sum_, product

my_sum, my_product = multireturn(3, 4)
print(my_sum) # 7
print(my_product) # 12
```



## Import functions from other files

```
import random

def random_add(a, b):
    return a + b + random.random()
```

Import the functions from myfunc.py in another file (just like we did with numpy):

```
from myfunc import random_add
print(random_add(1, 2))
```



# Fitting Data



#### Fitting Data

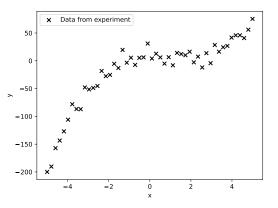
In physics we often have to fit data to a function. There are many ways to fit a function to given data. For now we will use <a href="mailto:numpy's polyfit">numpy's polyfit</a>() function.

It fits a polynomial of degree  $\deg$  to the data reducing the sum of squared residuals. It returns the coefficients of the polynomial in decreasing powers.



#### Fitting Data - Experimental Data

From our measurements we get the following data:



points.



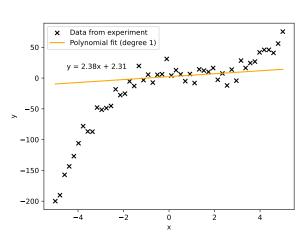
#### Fitting Data - Linear Fit

```
1 \times data = [...]
2 y_data = [...]
3 # plot data
4 plt.scatter(x_data, y_data, label="Data")
5 # Fit a line to the data (deg=1)
6 coefficients1 = np.polyfit(x_data, y_data,
    1)
7 # calculate the fitted values
8 y_fit1 = np.polyval(coefficients1, x_data)
9 plt.plot(x_data, y_fit1, label='Fit deg 1')
 polyval() evaluates the polynomial at the given
```



#### Fitting - Linear Fit

A linear fit on the data works well for  $x \in [-2,3]$ , but not outside this range.





### Fitting Data - Cubic Fit

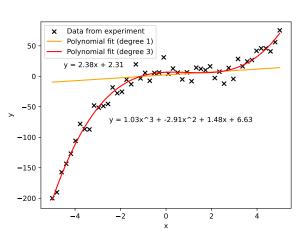
We can now add another fit, this time a cubic fit:

```
1 ...
2 # Fit a polynomial of degree 3 to the data
3 coefficients3 = np.polyfit(x_data, y_data,
3)
4 y_fit3 = np.polyval(coefficients3, x_data)
5 plt.plot(x_data, y_fit3, label='Fit deg 3')
```



#### Fitting - Cubic Fit

A cubic fit describes the data better than a linear fit.





### Fitting Data - Uncertainty

Every measurement has it's own uncertainty. We add uncertainty-bars to the data points using the <a href="mailto:errorbar()">errorbar()</a> function:

First two arguments are the data, yerr and xerr are the uncertainties. color specifies the color of the datapoints and ecolor the color of the errorbars.



#### Fitting Data - Uncertainty

