

01_Fundamentals

January 18, 2025

1 Modules/packages/libraries

Definitions:

- Modules: A module is a file which contains python functions, global variables etc. It is nothing but .py file which has python executable code / statement.
- Packages: A package is a namespace which contains multiple package/modules. It is a directory which contains a special file `__init__.py`
- Libraries: A library is a collection of various packages. There is no difference between package and python library conceptually.

Modules/packages/libraries can be easily “imported” and made functional in your python code. A set of libraries comes with every python installation. Others can be installed locally and then imported. Your own code sitting somewhere else in your local computer can be imported too.

Further details (very important!) on packages and how to create them can be found online. We may find the need of creating our own during the course.

1.1 notes from lecture

Jupyter stands from Ju-lia, Py-thon and R. these are the three programming languages that Jupyter was made for.

Modules is what i built up to this day.

cross-compatibility of libraries was a real issue with c++ libraries. In python with pip everything is checked automatically and you do not have to know and care about these things.

every know and then there might be an update to some library that breaks up cross-compatibility with other libraries. gith management of those libraries is a complicated thing.

```
[3]: ##### all the "stuff" that is in the math library can be used
import math
print(math.pi)

# you can give math a label for convenience
import math as m
print (m.pi)

# alternatively you can import only a given "thing" from the library
```

```

from math import pi    #you can add several libraries at once, just list them
    ↳separated by a ", "
print (pi)

# or just get everything (very dangerous!!!)
"""
dangerous because you do not know what is inside the library
never do this unless you know very well what is inside
for example, you can do it when you import a module that you wrote yourself
in math there may be variables and then you declare them again and you are
    ↳losing them
and you make a mess
"""
#from math import *
#print (sqrt(7))

```

```

3.141592653589793
3.141592653589793
3.141592653589793
2.6457513110645907

```

To know which modules are there for you to use just type:

```
[2]: print (help('modules') )
```

Please wait a moment while I gather a list of all available modules...

test_sqlite3: testing with SQLite version 3.42.0

2024-10-23 10:53:10.389 Python[1889:51019] WARNING: Secure coding is not enabled for restorable state! Enable secure coding by implementing NSApplicationDelegate.applicationSupportsSecureRestorableState: and returning YES.

IPython	_tracemalloc	heapq	reprlib
PIL	_typing	hmac	resource
__future__	_uuid	html	rlcompleter
__hello__	_warnings	http	runpy
__phello__	_weakref	idlelib	sched
_abc	_weakrefset	imaplib	scipy
_aix_support	_xxinterpchannels	imghdr	seaborn
_ast	_xxsubinterpreters	importlib	secrets
_asyncio	_xtestfuzz	inspect	select
_bisect	_zoneinfo	io	selectors
_blake2	abc	ipaddress	shelve
_bz2	aifc	ipykernel	shlex
_codecs	antigravity	ipykernel_launcher	shutil
_codecs_cn	appnope	itertools	signal
_codecs_hk	argparse	jedi	site

_codecs_iso2022	array	json	six
_codecs_jp	ast	jupyter	smtplib
_codecs_kr	asttokens	jupyter_client	sndhdr
_codecs_tw	asyncio	jupyter_core	socket
_collections	atexit	keyword	socketserver
_collections_abc	audioop	kiwisolver	sqlite3
_compat_pickle	base64	lib2to3	sre_compile
_compression	bdb	linecache	sre_constants
_contextvars	binascii	locale	sre_parse
_crypt	bisect	logging	ssl
_csv	builtins	lzma	stack_data
_ctypes	bz2	mailbox	stat
_ctypes_test	cProfile	mailcap	statistics
_curses	calendar	marshal	string
_curses_panel	cgi	math	stringprep
_datetime	cgilib	matplotlib	struct
_dbm	chunk	matplotlib_inline	subprocess
_decimal	cmath	mimetypes	sunau
_elementtree	cmd	mmap	symtable
_functools	code	modulefinder	sys
_hashlib	codecs	multiprocessing	sysconfig
_heapq	codeop	nest_asyncio	syslog
_imp	collections	netrc	tabnanny
_io	colorsys	nis	tarfile
_json	comm	nntplib	telnetlib
_locale	compileall	ntpath	tempfile
_lsprof	concurrent	nturl2path	termios
_lzma	configparser	numbers	test
_markupbase	contextlib	numpy	textwrap
_md5	contextvars	opcode	this
_multibytecodec	contourpy	operator	threading
_multiprocessing	copy	optparse	time
_opcode	copyreg	os	timeit
_operator	crypt	packaging	tkinter
_osx_support	csv	pandas	token
_pickle	ctypes	parso	tokenize
_posixshm	curses	pathlib	tomllib
_posixsubprocess	cycler	pdb	tornado
_py_abc	dataclasses	pexpect	tqdm
_pydatetime	datetime	pickle	trace
_pydecimal	dateutil	pickletools	traceback
_pyio	dbm	pip	tracemalloc
_pylong	debugpy	pipes	traitlets
_queue	decimal	pkgutil	tty
_random	decorator	platform	turtle
_scproxy	difflib	platformdirs	turtledemo
_sha1	dis	plistlib	types
_sha2	doctest	poplib	typing

_sha3	email	posix	tzdata
_signal	encodings	posixpath	unicodedata
_sitebuiltins	ensurepip	pprint	unittest
_socket	enum	profile	urllib
_sqlite3	errno	prompt_toolkit	uu
_sre	executing	pstats	uuid
_ssl	faulthandler	psutil	venv
_stat	fcntl	pty	warnings
_statistics	filecmp	ptyprocess	wave
_string	fileinput	pure_eval	wcwidth
_strptime	fnmatch	pwd	weakref
_struct	fontTools	py_compile	webbrowser
_symtable	fractions	pyclbr	wsgiref
_sysconfigdata__darwin_darwin	ftplib	pydoc	xdrlib
_testbuffer	functools	pydoc_data	xml
_testcapi	gc	pyexpat	xmlrpc
_testclinic	genericpath	pygments	xxlimited
_testimportmultiple	getopt	pylab	xxlimited_35
_testinternalcapi	getpass	pyparsing	xxsubtype
_testmultiphase	gettext	pytz	zipapp
_testsinglephase	glob	queue	zipfile
_thread	graphlib	quopri	zipimport
_threading_local	grp	random	zlib
_tkinter	gzip	re	zmq
_tokenize	hashlib	readline	zoneinfo

Enter any module name to get more help. Or, type "modules spam" to search for modules whose name or summary contain the string "spam".

None

/Users/miriamzara/LaboratoryOfComputationalPhysics_Y7/myenv/lib/python3.12/site-packages/tqdm/auto.py:21: TqdmWarning: IProgress not found. Please update jupyter and ipywidgets. See

https://ipywidgets.readthedocs.io/en/stable/user_install.html

```
from .autonotebook import tqdm as notebook_tqdm
```

pip is a special package. It is used from the command line to install properly (e.g. matching the version of the local packages) new packages. It can also be used from within python to check i.e. the set installed packages and their versions. N.B.: only the installed packages on top of the default ones will be listed

three numbers e.g. scipy==2.13.1

- *if the first number changes, it surely breaks compatibility*
- *the second one can break compatibility*
- *the third is used to mean fixing bugs*

```
[ ]: """
try:
    # this doesn't work anymore
    import pip
    sorted(["%s=%s" % (i.key, i.version) for i in pip._internal.utils.misc.
↳get_installed_distributions()])
    print ("within try")
"""

# use this instead
import pkg_resources
installed_packages = pkg_resources.working_set
installed_packages_list = sorted(["%s=%s" % (i.key, i.version)
    for i in installed_packages])
print(installed_packages_list)
```

2 Copies and Views

Copies:

- A **copy** creates a new object that is a duplicate of the original one.
- Changes made to the copy do not affect the original object, and vice versa.
- There are two types of copies:
 - **Shallow Copy**: Only the top-level object is copied. If the original object contains references to other objects (e.g., a list of lists), the references are copied, not the objects themselves.
 - **Deep Copy**: A complete copy is made, including all **nested** objects. Changes to any part of the deep copy won't affect the original object.

```
[5]: import copy
original = [1, 2, [3, 4]]
shallow_copy = copy.copy(original)
print("Original:", original) # Output: [1, 2, [3, 4]]

shallow_copy[0] = 100 # This won't affect the original
shallow_copy[2][0] = 300 # This will affect the original's nested list

print("Original:", original) # Output: [1, 2, [300, 4]]
print("Shallow Copy:", shallow_copy) # Output: [100, 2, [300, 4]]
```

```
Original: [1, 2, [3, 4]]
Original: [1, 2, [300, 4]]
Shallow Copy: [100, 2, [300, 4]]
```

Views:

- A view provides a **reference** to the original object without creating a new object.
- Changes made through the view directly affect the original object because the view is simply another way of accessing the original data.

- We will see in the next classes that NumPy arrays and **Pandas DataFrames** often deal with views when you slice or manipulate them.

this because when you have big data (GB or TB size) you really want to avoid duplicating as much as possible because you don't have that much memory

```
[ ]: import numpy as np
arr = np.array([1, 2, 3, 4])
view = arr[1:3] # This creates a view, not a copy

view[0] = 100 # Modifies the original array

print("Original array:", arr) # Output: [ 1 100  3  4]
print("View:", view) # Output: [100  3]
```

3 Functions

keys are valid when they can be HASHED (more or less: translated into a list of symbols)

for example, lists are hashable and as such they can be used as keys for a dictionary, but most of the times of course you use a string as key

```
[11]: def square(x):
        """Square of x."""
        return x*x

def cube(x):
    """Cube of x."""
    return x*x*x

# create a dictionary of functions
funcs = {
    'square': square,
    'cube': cube,
}

x = 3
print(square(x))
print(cube(x))

for func in sorted(funcs):
    print (func, funcs[func](x))
```

```
9
27
cube 27
square 9
```

3.1 Functions arguments

In Python, whether arguments passed to a function are treated as views or copies depends on the data type of the argument and how it is handled inside the function. Here's how **Python handles** various types of arguments:

Immutable Types (e.g., integers, strings, tuples): * For immutable types like `int`, `str`, and `tuple`, when you pass them as arguments to a function, Python creates a **copy** of the reference to the object, not the object itself. * Since these objects cannot be modified, **any attempt to change them inside the function results in the creation of a new object**, leaving the original object unchanged.

```
[8]: def modify(x):  
      x += 1  
      print("Inside function:", x)  
  
      a = 10  
      modify(a)  
      print("Outside function:", a)
```

```
Inside function: 11  
Outside function: 10
```

Mutable Types (e.g., lists, dictionaries, sets): * For mutable types like `list`, `dict`, or `set`, Python passes a reference to the original object. This means that any modification made to the object inside the function **will affect** the original object outside the function, as they both reference the same object (this behaves like a view).

```
[9]: def modify_list(lst):  
      lst.append(4)  
      print("Inside function:", lst)  
  
      my_list = [1, 2, 3]  
      modify_list(my_list)  
      print("Outside function:", my_list)
```

```
Inside function: [1, 2, 3, 4]  
Outside function: [1, 2, 3, 4]
```

3.2 Higher order functions

A function that uses another function as an input argument or returns a function is known as a higher-order function (HOF). The most familiar examples are `map` and `filter`.

*Also we look at **reduce** and **zip***

3.2.1 map

The map function applies a function to each member of a collection

```
[12]: x = list(map(square, range(5)))
      print (x)

      # Note the difference w.r.t python 2. In python 3 map returns an iterator so you
      ↪ can do stuff like:
      for i in map(square, range(5)): print(i)

      # or
      [i for i in map(square, range(6))]
```

[0, 1, 4, 9, 16]

0

1

4

9

16

```
[12]: [0, 1, 4, 9, 16, 25]
```

3.2.2 filter

The filter function applies a predicate to each member of a collection, **retaining only** those members where the predicate is True

```
[14]: def is_even(x):
      return x%2 == 0

      print (list(filter(is_even, range(5))))
      #print(type(range(5)))
```

[0, 2, 4]

<class 'range'>

Combinations in sequence of HOF are obviously possible

```
[ ]: list(map(square, filter(is_even, range(5))))
```

3.2.3 reduce

Not that common anymore but good to know because it was a big thing at the beginning of big data computing

The **reduce** function reduces a collection using a binary operator to combine items two at a time. More often than not reduce can be substituted with a more efficient for loop. It is worth mentioning it for its key role in big-data applications together with map (the map-reduce paradigm). N.B.: it no longer exists as a built-in function in python 3, it is now part of the **functools** library

```
[15]: from functools import reduce

      def my_add(x, y):
```



```

    return x + y

# another implementation of the sum function
reduce(my_add, [1,2,3,4,5]) #output = 15

```

[15]: 15

3.2.4 zip

zip is useful when you need to iterate over matched elements of multiple lists

```

[21]: xs = [1, 2, 3, 4]
      ys = [10, 20, 30, 40]
      zs = ['a', 'b', 'c', 'd']

      for x, y, z in zip(xs, ys, zs):
          print (x, y, z)

      print(list(zip(xs, ys, zs)))

      # No problem when sizes are not homogeneous
      # Python simply skips the elements with no pairing
      xs = [1, 2, 3]
      ys = [10, 20, 30, 40]
      zs = ['a', 'b', 'c',] # look there is an empty space and it doesnt matter
      print(list(zip(xs, ys, zs)))

```

```

1 10 a
2 20 b
3 30 c
4 40 d
[(1, 10, 'a'), (2, 20, 'b'), (3, 30, 'c'), (4, 40, 'd')]
[(1, 10, 'a'), (2, 20, 'b'), (3, 30, 'c')]

```

3.2.5 Custom HOF

```

[ ]: def custom_sum(xs, transform):
      """Returns the sum of xs after a user specified transform."""
      return sum(map(transform, xs))

      xs = range(5)
      print (custom_sum(xs, square))
      print (custom_sum(xs, cube))

```

3.2.6 Returning a function

```
[22]: # this is nice!!

def make_logger(target):
    def logger(data):
        with open(target, 'a') as f:
            f.write(data + '\n')
    return logger

foo_logger = make_logger('foo.txt') #foo.txt will be created if not there
↳ already
foo_logger('Hello')
foo_logger('World')
```

the exclamation mark ! at the beginning of the code cell

tells jupyter that what is next needs to be executed from the command line

cat is in fact a bash command

```
[24]: ! cat 'foo.txt'
```

Hello

World

3.3 Anonymous functions (lambda)

When using functional style, there is often the need to create specific functions that perform a limited task as input to a HOF such as map or filter. In such cases, these functions are often written as anonymous or lambda functions. The syntax is as follows:

lambda *arguments* : *expression*

If you find it hard to understand what a lambda function is doing, it should probably be rewritten as a regular function.

```
[31]: sum1 = lambda x,y: x+y
print(sum1(3,4), end= '\n\n')
print((lambda x,y: x+y)(3,4), end= '\n\n')
for i in map(lambda x: x*x, range(5)): print (i)
```

7

7

0

1

4

9
16

```
[32]: # what does this function do?
# answer:
# first make a list with [1, 4, 9, ... 100]
# then sum all the elements
from functools import reduce
s1 = reduce(lambda x, y: x+y, map(lambda x: x**2, range(1,10)))
print(s1)
```

285

3.4 Recursive functions

```
[33]: # nice !!!

def fib1(n):
    """Fib with recursion."""

    # base case
    if n==0 or n==1:
        return 1
    # recursive case
    else:
        return fib1(n-1) + fib1(n-2)

print ([fib1(i) for i in range(10)])
```

[1, 1, 2, 3, 5, 8, 13, 21, 34, 55]

```
[34]: # In Python, a more efficient version that does not use recursion is

def fib2(n):
    """Fib without recursion."""
    a, b = 0, 1
    for i in range(1, n+1):
        a, b = b, a+b # nice!
    return b

print ([fib2(i) for i in range(10)])
```

[1, 1, 2, 3, 5, 8, 13, 21, 34, 55]

```
[35]: # check indeed the timing:
# timeit runs the loop a high number of times, in order to provide you with an
↳ accurate estimate
%timeit fib1(20) # averaged over 1000 loops
```

```
%timeit fib2(20) # this is a thousand factor faster! averaged over 1,000,000
↳ loops
```

674 s \pm 1 s per loop (mean \pm std. dev. of 7 runs, 1,000 loops each)
359 ns \pm 1.2 ns per loop (mean \pm std. dev. of 7 runs, 1,000,000 loops each)

3.5 Iterators

for sake of completeness but we will never see them

Iterators represent streams of values. Because only one value is consumed at a time, they use very little memory. Use of iterators is very helpful for working with data sets too large to fit into RAM.

```
[ ]: # Iterators can be created from sequences with the built-in function iter()
```

```
xs = [1,2,3]
x_iter = iter(xs)

print (next(x_iter))
print (next(x_iter))
print (next(x_iter))
#print (next(x_iter))
```

```
[ ]: # Most commonly, iterators are used (automatically) within a for loop
# which terminates when it encounters a StopIteration exception
```

```
x_iter = iter(xs)
for x in x_iter:
    print (x)
```

3.6 More on comprehensions

```
[ ]: # A generator expression
```

```
print ((x for x in range(10)))
```

```
# A list comprehension
```

```
print ([x**2 for x in range(10)])
```

```
# A set comprehension
```

```
print ({x for x in range(10)})
```

```
# A dictionary comprehension
```

```
print ({x: "x" for x in range(10) if x%2==0})
```

3.7 Useful Modules

nowadays this stuff is already inside high level functions of libraries like pandas or numpy

these native modules are not something we will use, but here just for completeness

You may want to have a look at the content of the following modules for further usage of (HO) functions: - [operator](#) - [functools](#) - [itertools](#) - [toolz](#) - [functools](#)

3.8 Decorators

Decorators are a type of HOF that take a function and return a wrapped function that provides additional useful properties.

Examples:

- logging
- profiling
- Just-In-Time (JIT) compilation

..

- logging: you may want to record things happening while the functions is executed and store in log file
- profiling: almost same thing

```
[38]: def my_decorator(func):
      def wrapper():
          print("Something is happening before the function is called.")
          func()
          print("Something is happening after the function is called.")
      return wrapper

      def say_whee():
          print("Whee!")

      say_whee = my_decorator(say_whee) # say_whee is now the decorated function
```

```
[39]: say_whee()
```

Something is happening before the function is called.

Whee!

Something is happening after the function is called.

Python allows you to use decorators in a **simpler way** with the @ symbol, sometimes called the “pie” syntax

```
[40]: def my_decorator(func):
      def wrapper():
          print("Something is happening before the function is called.")
          func()
          print("Something is happening after the function is called.")
```

```

    return wrapper

@my_decorator
def say_whee():
    print("Whee!")

```

```
[41]: say_whee()
```

Something is happening before the function is called.

Whee!

Something is happening after the function is called.

JIT A JIT (Just-In-Time) compiler refers to a technique used to **improve the performance** of code execution by **compiling** code into machine code at runtime, rather than interpreting it line by line. This helps speed up program execution by reducing the overhead of repeatedly interpreting code.

While the standard Python interpreter, CPython, does not include a JIT compiler, there are alternative Python implementations that provide JIT compilation, such as PyPy or [numba](#).

JIT combines both interpretation and compilation. It interprets code initially, and as it identifies sections of code that are executed repeatedly, it compiles them to machine code, optimizing performance dynamically.

remember! python is slow because it is not compiled, it is interpreted.

```
[ ]: from numba import jit

@jit
def fast_sum(arr):
    total = 0
    for i in arr:
        total += i
    return total

```

```
[ ]: fast_sum(list(range(100)))
```

```
# is it really fast? Try to compare with my_add above
```

4 Classes and Objects

classes and objects are obsolete now (since 20 years). we do not need to have our main focus on this anymore. Nowadays our main focus is manipulation of enormous datasets in a distributed environment. this is what unabled AI to ramp up. Object-oriented programming is still here and still used, but nowadays the crucial things are others

Old school object-oriented programming is possible and often used in python. Classes are defined similarly to standard object-oriented languages, with similar functionalities.

The main python doc [page](#) is worth reading through

```
[42]: class Pet:
    # the "constructor"
    def __init__(self, name, age): #initalize the elements of the class
        self.name=name
        self.age=age
    # class functions take the "self" parameter !!!
    def set_name(self,name):
        self.name=name
    def convert_age(self,factor):
        self.age*=factor

buddy=Pet("buddy",4)
print (buddy.name, buddy.age)
buddy.age=3
print (buddy.convert_age(4))
print (buddy.age)
```

```
buddy 4
None
12
```

```
[43]: # inheritance is straightforward

# new thing to know!!

class Dog(Pet):
    # the following variables is "global", i.e. holds for all "Dog" objects
    species = "mammal"
    # functions can be redefined as usual
    def convert_age(self):
        self.age*=7
    def set_species(self, species):
        self.species = species

puppy=Dog("tobia",10)
print(puppy.name)
puppy.convert_age()
print(puppy.age)
```

```
tobia
70
```