

# Section 5 Class and Modules

A possibly overlooked point: Modules and Class in Python share many similarities at the basic level. They both contain some data (names, attributes) and codes (functions, methods) for the convenience of users -- and the codes to call them are also similar. Of course, Class also serves as the blue prints to generate instances, and supports more advanced functions such as Inheritance.

## Class and Instance

Intuitively speaking, **classes** (or understood as types) are the "factories" to produce **instances** (concrete objects). For example, you can image that in the class of "list" in python, it defines the behavior of lists (methods) such as `append` , `copy` , and you can create concrete list objects (each with different values) from the list class, and directly uses the methods defined.

Programming with the idea of creating classes is the key to [Object-Oriented Programming\(OOP\)](#).

## Simple Example of Vector

Let's first define the simplest class in Python

```
In [6]: class VectorV0:
        '''The simplest class in python''' # this is the document string

        pass
```

and create two instances `v1` and `v2`

```
In [8]: v1 = VectorV0() # note the parentheses here, they are the grammar to create instance f
print(id(v1))
v2 = VectorV0()
print(id(v2))
```

2081447241328  
2081447241376

Now `v1` and `v2` are the objects in Python

```
In [3]: type(v1)
```

```
Out[3]: __main__.VectorV0
```

```
In [4]: dir(v1)
```

```
Out[4]: ['_class_',
          '_delattr_',
          '_dict_',
          '_dir_',
          '_doc_',
          '_eq_',
```

```

'__format__',
'__ge__',
'__getattr__',
'__gt__',
'__hash__',
'__init__',
'__init_subclass__',
'__le__',
'__lt__',
'__module__',
'__ne__',
'__new__',
'__reduce__',
'__reduce_ex__',
'__repr__',
'__setattr__',
'__sizeof__',
'__str__',
'__subclasshook__',
'__weakref__']

```

In [9]:

```
help(v1)
```

Help on VectorV0 in module \_\_main\_\_ object:

```

class VectorV0(builtins.object)
|   The simplest class in python
|
|   Data descriptors defined here:
|
|   __dict__
|       dictionary for instance variables (if defined)
|
|   __weakref__
|       list of weak references to the object (if defined)

```

We can manually assign the attributes to instance v1 and v2

In [13]:

```

import math

v1.x = 1.0 # this is called instance attributes
v1.y = 2.0
v1.norm = math.sqrt(5)

v2.x = 2.0
v2.y = 3.0

```

In [14]:

```
dir(v1)
```

Out[14]:

```

['__class__',
'__delattr__',
'__dict__',
'__dir__',
'__doc__',
'__eq__',
'__format__',
'__ge__',
'__getattr__',
'__gt__',
'__hash__',

```

```

'__init__',
'__init_subclass__',
'__le__',
'__lt__',
'__module__',
'__ne__',
'__new__',
'__reduce__',
'__reduce_ex__',
'__repr__',
'__setattr__',
'__sizeof__',
'__str__',
'__subclasshook__',
'__weakref__',
'norm',
'x',
'y']

```

We don't want to create the instance or define the coordinates separately. Can we do these in one step, when initializing the instance?

```

In [16]: class VectorV1:
        '''define the vector''' # this is the document string
        dim = 2 # this is the attribute in class -- class attributes
        def __init__(self, x=0.0, y=0.0): # any method in Class requires the first paramet
            self.x = x # instance attributes defined by self.attr
            self.y = y

```

```

In [19]: v1 = VectorV1(4.0,2.0) # you can pass the value directly, because you defined the __ini

```

```

In [21]: dir(v1)

```

```

Out[21]: ['__class__',
'__delattr__',
'__dict__',
'__dir__',
'__doc__',
'__eq__',
'__format__',
'__ge__',
'__getattr__',
'__gt__',
'__hash__',
'__init__',
'__init_subclass__',
'__le__',
'__lt__',
'__module__',
'__ne__',
'__new__',
'__reduce__',
'__reduce_ex__',
'__repr__',
'__setattr__',
'__sizeof__',
'__str__',
'__subclasshook__',
'__weakref__',
'dim',

```

```
'x',  
'y']
```

In [22]:

```
print(v1.dim)  
print(v1.x)  
print(v1.y)
```

```
2  
4.0  
2.0
```

Btw, there is nothing mysterious about the `__init__` : you can just assume it is a function (method) stored in `v1`, and you can always call it if you like!

When you write `v1.__init__()` , you can equivalently think that you are calling a function with "ugly function name" `__init__` , and the parameter is `v1` (self), i.e. you are writing `__init__(v1)` . It is just a function updating the attributes of instance objects!

More generally, for the method `method(self, params)` you can call it by `self.method(params)` .

In [24]:

```
print(v1.x)  
print(id(v1))  
y = v1.__init__(2,7) #reinitializes the values of our vector  
print(v1.x)  
print(id(v1))  
print(y)
```

```
0.0  
2081447541296  
2  
2081447541296  
None
```

`v1` is just like a mutable object, and the "function" `__init__( )` just change `v1` in place!

Now we move on to update our vector class by defining more functions. Since you may not like ugly names here with dunder (a.k.a **double underscore**), let's just begin with normal function names.

In [26]:

```
import math  
  
class VectorV2:  
    '''define the vector''' # this is the document string  
    dim = 2 # this is the class attribute  
  
    def __init__(self, x=0.0, y=0.0): # any method in Class requires the first paramet  
        '''initialize the vector by providing x and y coordinate'''  
        self.x = x  
        self.y = y  
  
    def norm(self):  
        '''calculate the norm of vector'''  
        return math.sqrt(self.x**2+self.y**2)  
  
    def vector_sum(self, other):  
        '''calculate the vector sum of two vectors'''  
        return VectorV2(self.x + other.x, self.y + other.y)
```

```
def show_coordinate(self):
    '''display the coordinates of the vector'''
    return 'Vector(%r, %r)' % (self.x, self.y)
```

In [27]:

```
help(VectorV2)
```

Help on class VectorV2 in module \_\_main\_\_:

```
class VectorV2(builtins.object)
|   VectorV2(x=0.0, y=0.0)
|
|   define the vector
|
|   Methods defined here:
|
|   __init__(self, x=0.0, y=0.0)
|       initialize the vector by providing x and y coordinate
|
|   norm(self)
|       calculate the norm of vector
|
|   show_coordinate(self)
|       display the coordinates of the vector
|
|   vector_sum(self, other)
|       calculate the vector sum of two vectors
|
|   -----
|   Data descriptors defined here:
|
|   __dict__
|       dictionary for instance variables (if defined)
|
|   __weakref__
|       list of weak references to the object (if defined)
|
|   -----
|   Data and other attributes defined here:
|
|   dim = 2
```

In [37]:

```
v1 = VectorV2(1.0, 2.0)
v2 = VectorV2(2.0, 3.0)
print(v1.x)
```

1.0

In [32]:

```
v1_length = v1.norm()
print(v1_length)
```

2.23606797749979

Equivalent way to call this method is (although not used often):

In [31]:

```
VectorV2.norm(v1) #calling the method through the class name rather than the object nam
```

Out[31]: 2.23606797749979

Even for built-in types, we have something similar

```
In [33]: a = [1,2,3]
list.append(a,4) # equivalent to a.append(4), note that list is the class name
print(a)
a.append(9372)
print(a)
```

```
[1, 2, 3, 4]
[1, 2, 3, 4, 9372]
```

despite that we don't have any reason not to use `a.append()` directly.

```
In [34]: v3 = v1.vector_sum(v2)
v3.show_coordinate()
```

```
Out[34]: 'Vector(3.0, 5.0)'
```

```
In [35]: v1+v2 # will it work?
```

```
-----
TypeError                                 Traceback (most recent call last)
<ipython-input-35-3b50698d06d4> in <module>
----> 1 v1+v2 # will it work?

TypeError: unsupported operand type(s) for +: 'VectorV2' and 'VectorV2'
```

```
In [36]: print(v3)
v3
```

```
<__main__.VectorV2 object at 0x000001E49FF0FB50>
```

```
Out[36]: <__main__.VectorV2 at 0x1e49ff0fb50>
```

Something that we are still not satisfied:

- By typing `v3` or using `print()` in the code, we cannot show its coordinates directly
- We cannot use the `+` operator to calculate the vector sum

## Special (Magic) Methods

Here's the magic: by merely changing the function name, we can realize our goal!

```
In [38]: class VectorV3:
        '''define the vector''' # this is the document string
        dim = 2 # this is the attribute

        def __init__(self, x=0.0, y=0.0): # any method in Class requires the first paramet
        '''initialize the vector by providing x and y coordinate'''
            self.x = x
            self.y = y

        def norm(self):
            '''calculate the norm of vector'''
            return math.sqrt(self.x**2+self.y**2)
```

```
def __add__(self, other):
    '''calculate the vector sum of two vectors'''
    return VectorV3(self.x + other.x, self.y + other.y)

def __repr__(self): #special method of string representation
    '''display the coordinates of the vector'''
    return 'Vector(%r, %r)' % (self.x, self.y)
```

In [39]:

```
help(VectorV3)
```

Help on class VectorV3 in module \_\_main\_\_:

```
class VectorV3(builtins.object)
|   VectorV3(x=0.0, y=0.0)
|
|   define the vector
|
|   Methods defined here:
|
|   __add__(self, other)
|       calculate the vector sum of two vectors
|
|   __init__(self, x=0.0, y=0.0)
|       initialize the vector by providing x and y coordinate
|
|   __repr__(self)
|       display the coordinates of the vector
|
|   norm(self)
|       calculate the norm of vector
|
|   -----
|   Data descriptors defined here:
|
|   __dict__
|       dictionary for instance variables (if defined)
|
|   __weakref__
|       list of weak references to the object (if defined)
|
|   -----
|   Data and other attributes defined here:
|
|   dim = 2
```

In [40]:

```
v1 = VectorV3(1.0,2.0)
v2 = VectorV3(2.0,3.0)
```

In [41]:

```
v3 = v1.__add__(v2) # just call special methods as ordinary methods
v3.__repr__()
```

Out[41]: 'Vector(3.0, 5.0)'

In [45]:

```
v3 = v1+v2 # here is the point of using special methods!
```

In [46]:

```
print(v3)
```

```
Vector(3.0, 5.0)
```

Special methods are just like VIP admissions to take full use of the built-in operators in Python. With other special methods (such as for container methods), you can even get elements by index `v3[0]` , or iterate through the object you created. For more advanced usage, you can [see here](#).

```
In [ ]: #Exercise: What special method is required to define subtraction?
```

## (Optional) More Comments about `__repr__()` and `__str__()`

These are all the methods to display some strings about the object. An obvious difference is that when you directly **run** (evaluate) the object in code cell, it will execute `__repr__` , and when you **print** the object, it will first execute `__str__` . If `__str__` is not defined, then when calling `print` , the `__repr__` will be executed, but not vice versa. For more information, see the discussion [here](#).

```
In [47]: class VectorV3_1:
        '''define the vector''' # this is the document string
        dim = 2 # this is the attribute

        def __init__(self, x=0.0, y=0.0): # any method in Class requires the first paramet
        '''initialize the vector by providing x and y coordinate'''
            self.x = x
            self.y = y

        def __repr__(self): #special method of string representation
        '''display the coordinates of the vector'''
            return 'repr: Vector(%r, %r)' % (self.x, self.y)

        def __str__(self): #special method of string representation
        '''display the coordinates of the vector'''
            return 'str: vector[%r, %r]' % (self.x, self.y)
```

```
In [48]: v1 = VectorV3_1(1.0,2.0)
```

```
In [49]: v1 # directly call in cell code, or from repr() function
```

```
Out[49]: repr: Vector(1.0, 2.0)
```

```
In [40]: print(v1)
```

```
str: vector[1.0, 2.0]
```

## Inheritance

Now we want to add another scalar production method to Vector, but we're tired of rewriting all the other methods. A good way is to create new Class VectorV4 (Child Class) by inheriting from VectorV3 (Parent Class) that we have already defined.



```
In [51]: class VectorV4(VectorV3): # Note the class VectorV3 in parentheses here
        '''define the vector''' # this is the document string
        def __mul__(self, scalar):
            '''calculate the scalar product'''
            return VectorV4(self.x * scalar, self.y * scalar)
```

```
In [53]: help(VectorV4)
```

Help on class VectorV4 in module \_\_main\_\_:

```
class VectorV4(VectorV3)
  VectorV4(x=0.0, y=0.0)

  define the vector

  Method resolution order:
    VectorV4
    VectorV3
    builtins.object

  Methods defined here:

    __mul__(self, scalar)
        calculate the scalar product

  -----
  Methods inherited from VectorV3:

    __add__(self, other)
        calculate the vector sum of two vectors

    __init__(self, x=0.0, y=0.0)
        initialize the vector by providing x and y coordinate

    __repr__(self)
        display the coordinates of the vector

    norm(self)
        calculate the norm of vector

  -----
  Data descriptors inherited from VectorV3:

    __dict__
        dictionary for instance variables (if defined)

    __weakref__
        list of weak references to the object (if defined)

  -----
  Data and other attributes inherited from VectorV3:

    dim = 2
```

```
In [54]: v1 = VectorV4(1.0,2.0)
        v2 = VectorV4(2.0,3.0)
```

```
In [55]: v1+v2
```

```
Out[55]: Vector(3.0, 5.0)
```

```
In [57]: v1*20
```

```
Out[57]: Vector(20.0, 40.0)
```

## Modules and Packages

In Python, Functions (plus Classes, Variables) are contained in Modules, and Modules are organized in directories of Packages. In fact, Modules are also objects in Python!

Now we have the `Vector.py` file in the folder. When we import the module, the interpreter will create a name `Vector` pointing to the module object. The functions/classes/variables defined in the module can be called with `Vector.XXX`, i.e. they are in the **namespace** of `Vector` (can be seen through `dir`).

Of course, the (annoying) rules of object assignment (be careful about changing mutable objects even in modules) in Python still applies, but we won't go deep in this course.

```
In [60]: import Vector
print(type(Vector))
dir(Vector) # 'attributes' (namespace) in the module Vector -- note the variables/funct
```

```
Out[60]: <class 'module'>
['VectorV5',
 '__builtins__',
 '__cached__',
 '__doc__',
 '__file__',
 '__loader__',
 '__name__',
 '__package__',
 '__spec__',
 'print_hello',
 'string']
```

```
In [61]: string
```

```
-----
NameError                                Traceback (most recent call last)
<ipython-input-61-edbf08a562d5> in <module>
----> 1 string

NameError: name 'string' is not defined
```

```
In [66]: Vector.string
```

```
Out[66]: 'Python'
```

```
In [63]: Vector.print_hello()
```

```
Hello
```

```
In [64]: v5 = Vector.VectorV5(1.0,2.0)
v5
```

```
Out[64]: Vector(1.0, 2.0)
```

Other different ways to import module:

```
In [65]: import Vector as vc # create a name vc point to the module Vector.py -- good practice,
vc.string
```

```
Out[65]: 'Python'
```

```
In [71]: from Vector import print_hello # may cause some name conflicts if write larger programs
print_hello() # Where does this print_hello come from ? It may take some time to figure
```

```
import VectorCopy
Vector.print_hello()
VectorCopy.print_hello()
```

```
Hello
Hello
Sup
```

It's totally possible that different modules (packages) contain same names. Some problems may happen if we try the from...import way. That's why the first way (import or import as) is always recommended.

```
In [72]: import math
import numpy as np
print(math.cos(math.pi))# everything is clear -- there won't be any confusions
print(np.cos(np.pi))# everything is clear -- there won't be any confusions
```

```
-1.0
-1.0
```

```
In [55]: from Vector import * # Be careful about import everything -- may cause serious name con
string
```

```
Out[55]: 'Python'
```

To import the modules, you must ensure that they are in your system paths.

```
In [73]: import sys
sys.path
```

```
Out[73]: ['C:\\Users\\Luke\\Math_10_SS1',
'E:\\ProgramData\\Anaconda3\\python38.zip',
'E:\\ProgramData\\Anaconda3\\DLLs',
'E:\\ProgramData\\Anaconda3\\lib',
'E:\\ProgramData\\Anaconda3',
'',
'E:\\ProgramData\\Anaconda3\\lib\\site-packages',
'E:\\ProgramData\\Anaconda3\\lib\\site-packages\\loket-0.2.1-py3.8.egg',
'E:\\ProgramData\\Anaconda3\\lib\\site-packages\\win32',
'E:\\ProgramData\\Anaconda3\\lib\\site-packages\\win32\\lib',
```

```
'E:\\ProgramData\\Anaconda3\\lib\\site-packages\\Pythonwin',  
'E:\\ProgramData\\Anaconda3\\lib\\site-packages\\IPython\\extensions',  
'C:\\Users\\Luke\\.ipython']
```

```
In [ ]: sys.modules.keys() # check all the modules are currently imported in the kernel
```

We can import the `inspect` module and use `getsource` function to see the source codes of imported modules.

```
In [75]: import inspect # this inspect itself is a module!  
lines = inspect.getsource(Vector.VectorV5)  
print(lines)  
  
class VectorV5:  
    '''define the vector''' # this is the document string  
    dim = 2 # this is the attribute  
  
    def __init__(self, x=0.0, y=0.0): # any method in Class requires the first parameter to be self!  
        '''initialize the vector by providing x and y coordinate'''  
        self.x = x  
        self.y = y  
  
    def norm(self):  
        '''calculate the norm of vector'''  
        return math.sqrt(self.x**2+self.y**2)  
  
    def __add__(self, other):  
        '''calculate the vector sum of two vectors'''  
        return VectorV5(self.x + other.x, self.y + other.y)  
  
    def __repr__(self): #special method of string representation  
        '''display the coordinates of the vector'''  
        return 'Vector(%r, %r)' % (self.x, self.y)  
  
    def __mul__(self, scalar):  
        '''calculate the scalar product'''  
        return VectorV5(self.x * scalar, self.y * scalar)
```

Note that this does not work for some Python modules/functions (Because they are written in C language).

You can view all the source codes of Python [here](#). Here is the complete documentation for reference about [standard Python library](#) -- the .py files that are now in your computer when you install python!

```
In [77]: import math # this won't work, because math is the built-in function -- written in C language!  
lines = inspect.getsource(math.sqrt) # will print the error  
print(lines)
```

```
-----  
TypeError                                 Traceback (most recent call last)  
<ipython-input-77-c1fae04687db> in <module>  
      1 import math # this won't work, because math is the built-in function -- written  
      in C language!  
----> 2 lines = inspect.getsource(math.sqrt) # will print the error  
      3 print(lines)  
  
E:\\ProgramData\\Anaconda3\\lib\\inspect.py in getsource(object)
```

```

995     or code object. The source code is returned as a single string. An
996     OSError is raised if the source code cannot be retrieved."""
--> 997     lines, lnum = getsourcelines(object)
998     return ''.join(lines)
999

E:\ProgramData\Anaconda3\lib\inspect.py in getsourcelines(object)
977     raised if the source code cannot be retrieved."""
978     object = unwrap(object)
--> 979     lines, lnum = findsourcelines(object)
980
981     if istraceback(object):

E:\ProgramData\Anaconda3\lib\inspect.py in findsourcelines(object)
778     is raised if the source code cannot be retrieved."""
779
--> 780     file = getsourcefile(object)
781     if file:
782         # Invalidate cache if needed.

E:\ProgramData\Anaconda3\lib\inspect.py in getsourcefile(object)
694     Return None if no way can be identified to get the source.
695     """
--> 696     filename = getfile(object)
697     all_bytecode_suffixes = importlib.machinery.DEBUG_BYTECODE_SUFFIXES[:]
698     all_bytecode_suffixes += importlib.machinery.OPTIMIZED_BYTECODE_SUFFIXES[:]

E:\ProgramData\Anaconda3\lib\inspect.py in getfile(object)
674     if iscode(object):
675         return object.co_filename
--> 676     raise TypeError('module, class, method, function, traceback, frame, or '
677                       'code object was expected, got {}'.format(
678                           type(object).__name__))

TypeError: module, class, method, function, traceback, frame, or code object was expected, got builtin_function_or_method

```

In [78]:

```

import copy # this can work, because copy.py is the "lib" folder and is written in Python
lines = inspect.getsource(copy.deepcopy) # no problem
print(lines)

```

```

def deepcopy(x, memo=None, _nil=[]):
    """Deep copy operation on arbitrary Python objects.

    See the module's __doc__ string for more info.
    """

    if memo is None:
        memo = {}

    d = id(x)
    y = memo.get(d, _nil)
    if y is not _nil:
        return y

    cls = type(x)

    copier = _deepcopy_dispatch.get(cls)
    if copier is not None:
        y = copier(x, memo)
    else:
        if issubclass(cls, type):
            y = _deepcopy_atomic(x, memo)
        else:

```

```

copier = getattr(x, "__deepcopy__", None)
if copier is not None:
    y = copier(memo)
else:
    reducer = dispatch_table.get(cls)
    if reducer:
        rv = reducer(x)
    else:
        reducer = getattr(x, "__reduce_ex__", None)
        if reducer is not None:
            rv = reducer(4)
        else:
            reducer = getattr(x, "__reduce__", None)
            if reducer:
                rv = reducer()
            else:
                raise Error(
                    "un(deep)copyable object of type %s" % cls)
    if isinstance(rv, str):
        y = x
    else:
        y = _reconstruct(x, memo, *rv)

# If is its own copy, don't memoize.
if y is not x:
    memo[d] = y
    _keep_alive(x, memo) # Make sure x lives at least as long as d
return y

```

```
In [79]: inspect.getsourcefile(copy) # see? the copy.py is in our local computer
```

```
Out[79]: 'E:\\ProgramData\\Anaconda3\\lib\\copy.py'
```

## Notes on Numpy Package

If we are interested in `numpy` that we're going to talk about in details soon -- in fact `numpy` is a package rather than modules. Package can contain many modules (some are also called subpackages, their difference is not important for our course) -- for example, the module (or subpackage, which is in the sub-directory of `numpy`) of [linalg](#).

```
In [ ]: import numpy as np # import the package numpy, and assign the "nickname" np to it
        [name for name in sys.modules.keys() if name.startswith('numpy')] # check what modules
```

```
In [ ]: print(np)
        dir(np) # namespace of numpy package -- it also includes the functions in np.core
```

Something special about `numpy`: The namespace of `numpy` contains both modules (e.g. `linalg` module) and functions (e.g. `sum` function). In fact, these functions are imported from the modules (subpackages) `numpy.core` or `numpy.lib` -- they are loaded only for the convenience of users, because of their high frequency in usage. For a more complete understanding, we can go to see the structure of `numpy` package in [GitHub](#).

```
In [82]: type(np.linalg)
```

Out[82]: module

```
In [83]: type(np.sum)
```

Out[83]: function

```
In [84]: print(id(np.core.sum))
print(id(np.sum))# see? np.sum is the same function with np.core.sum. In your usage, pl
np.core.sum is np.sum
```

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Out[84]: True

```
In [85]: print(inspect.getsource(np.sum))# Let's see the source code of sum function
```

```
@array_function_dispatch(_sum_dispatcher)
def sum(a, axis=None, dtype=None, out=None, keepdims=np._NoValue,
        initial=np._NoValue, where=np._NoValue):
    """
    Sum of array elements over a given axis.

    Parameters
    -----
    a : array_like
        Elements to sum.
    axis : None or int or tuple of ints, optional
        Axis or axes along which a sum is performed. The default,
        axis=None, will sum all of the elements of the input array. If
        axis is negative it counts from the last to the first axis.

        .. versionadded:: 1.7.0

    If axis is a tuple of ints, a sum is performed on all of the axes
    specified in the tuple instead of a single axis or all the axes as
    before.
    dtype : dtype, optional
        The type of the returned array and of the accumulator in which the
        elements are summed. The dtype of `a` is used by default unless `a`
        has an integer dtype of less precision than the default platform
        integer. In that case, if `a` is signed then the platform integer
        is used while if `a` is unsigned then an unsigned integer of the
        same precision as the platform integer is used.
    out : ndarray, optional
        Alternative output array in which to place the result. It must have
        the same shape as the expected output, but the type of the output
        values will be cast if necessary.
    keepdims : bool, optional
        If this is set to True, the axes which are reduced are left
        in the result as dimensions with size one. With this option,
        the result will broadcast correctly against the input array.

        If the default value is passed, then `keepdims` will not be
        passed through to the `sum` method of sub-classes of
        `ndarray`, however any non-default value will be. If the
        sub-class' method does not implement `keepdims` any
        exceptions will be raised.
    initial : scalar, optional
        Starting value for the sum. See `~numpy.ufunc.reduce` for details.
```

.. versionadded:: 1.15.0

where : array\_like of bool, optional  
Elements to include in the sum. See `~numpy.ufunc.reduce` for details.

.. versionadded:: 1.17.0

Returns

-----

sum\_along\_axis : ndarray

An array with the same shape as `a`, with the specified axis removed. If `a` is a 0-d array, or if `axis` is None, a scalar is returned. If an output array is specified, a reference to `out` is returned.

See Also

-----

ndarray.sum : Equivalent method.

add.reduce : Equivalent functionality of `add`.

cumsum : Cumulative sum of array elements.

trapz : Integration of array values using the composite trapezoidal rule.

mean, average

Notes

-----

Arithmetic is modular when using integer types, and no error is raised on overflow.

The sum of an empty array is the neutral element 0:

```
>>> np.sum([])
0.0
```

For floating point numbers the numerical precision of sum (and `np.add.reduce`) is in general limited by directly adding each number individually to the result causing rounding errors in every step. However, often numpy will use a numerically better approach (partial pairwise summation) leading to improved precision in many use-cases. This improved precision is always provided when no `axis` is given. When `axis` is given, it will depend on which axis is summed. Technically, to provide the best speed possible, the improved precision is only used when the summation is along the fast axis in memory. Note that the exact precision may vary depending on other parameters. In contrast to NumPy, Python's `math.fsum` function uses a slower but more precise approach to summation. Especially when summing a large number of lower precision floating point numbers, such as `float32`, numerical errors can become significant. In such cases it can be advisable to use `dtype="float64"` to use a higher precision for the output.

Examples

-----

```
>>> np.sum([0.5, 1.5])
2.0
>>> np.sum([0.5, 0.7, 0.2, 1.5], dtype=np.int32)
1
>>> np.sum([[0, 1], [0, 5]])
6
>>> np.sum([[0, 1], [0, 5]], axis=0)
array([0, 6])
```



```
>>> np.sum([[0, 1], [0, 5]], axis=1)
array([1, 5])
>>> np.sum([[0, 1], [np.nan, 5]], where=[False, True], axis=1)
array([1., 5.])
```

If the accumulator is too small, overflow occurs:

```
>>> np.ones(128, dtype=np.int8).sum(dtype=np.int8)
-128
```

You can also start the sum with a value other than zero:

```
>>> np.sum([10], initial=5)
15
"""
if isinstance(a, _gentype):
    # 2018-02-25, 1.15.0
    warnings.warn(
        "Calling np.sum(generator) is deprecated, and in the future will give a different result. "
        "Use np.sum(np.fromiter(generator)) or the python sum builtin instead.",
        DeprecationWarning, stacklevel=3)

    res = _sum_(a)
    if out is not None:
        out[...] = res
    return out
return res

return _wrapreduction(a, np.add, 'sum', axis, dtype, out, keepdims=keepdims,
                      initial=initial, where=where)
```

```
In [86]: 'eig' in dir(np) # where is the eigen value/vector function?
```

```
Out[86]: False
```

```
In [87]: np.eig # Won't work! Because eig is not defined in numpy (core) module!
```

```
-----
AttributeError                                Traceback (most recent call last)
<ipython-input-87-a5400bd55fe7> in <module>
----> 1 np.eig # Won't work! Because eig is not defined in numpy (core) module!

E:\ProgramData\Anaconda3\lib\site-packages\numpy\__init__.py in __getattr__(attr)
    301         return Tester
    302
--> 303         raise AttributeError("module {!r} has no attribute "
    304                             "{!r}".format(__name__, attr))
    305

AttributeError: module 'numpy' has no attribute 'eig'
```

```
In [88]: print(np.linalg) # np.linalg is a module(subpackage) -- its namespace containing many f
dir(np.linalg) # Let's check the names (functions) in linalg
```

```
<module 'numpy.linalg' from 'E:\\ProgramData\\Anaconda3\\lib\\site-packages\\numpy\\linalg\\__init__.py'>
```

```
Out[88]: ['LinAlgError',
          '__builtins__',
```

```

'__cached__',
'__doc__',
'__file__',
'__loader__',
'__name__',
'__package__',
'__path__',
'__spec__',
'_umath_linalg',
'cholesky',
'cond',
'det',
'eig',
'eigh',
'eigvals',
'eigvalsh',
'inv',
'lapack_lite',
'linalg',
'lstsq',
'matrix_power',
'matrix_rank',
'multi_dot',
'norm',
'pinv',
'qr',
'slogdet',
'solve',
'svd',
'tensorinv',
'tensorsolve',
'test']

```

```
In [89]: help(np.linalg.eig) # eig function is here! Don't forget to import numpy as np first
```

Help on function eig in module numpy.linalg:

```

eig(a)
    Compute the eigenvalues and right eigenvectors of a square array.

    Parameters
    -----
    a : (... , M, M) array
        Matrices for which the eigenvalues and right eigenvectors will
        be computed

    Returns
    -----
    w : (... , M) array
        The eigenvalues, each repeated according to its multiplicity.
        The eigenvalues are not necessarily ordered. The resulting
        array will be of complex type, unless the imaginary part is
        zero in which case it will be cast to a real type. When `a`
        is real the resulting eigenvalues will be real (0 imaginary
        part) or occur in conjugate pairs

    v : (... , M, M) array
        The normalized (unit "length") eigenvectors, such that the
        column ``v[:,i]`` is the eigenvector corresponding to the
        eigenvalue ``w[i]``.

    Raises
    -----
    LinAlgError

```

If the eigenvalue computation does not converge.

See Also

-----  
eigvals : eigenvalues of a non-symmetric array.  
eigh : eigenvalues and eigenvectors of a real symmetric or complex  
Hermitian (conjugate symmetric) array.  
eigvalsh : eigenvalues of a real symmetric or complex Hermitian  
(conjugate symmetric) array.  
scipy.linalg.eig : Similar function in SciPy that also solves the  
generalized eigenvalue problem.  
scipy.linalg.schur : Best choice for unitary and other non-Hermitian  
normal matrices.

Notes

-----

.. versionadded:: 1.8.0

Broadcasting rules apply, see the `numpy.linalg` documentation for details.

This is implemented using the ```_geev``` LAPACK routines which compute the eigenvalues and eigenvectors of general square arrays.

The number `w` is an eigenvalue of `a` if there exists a vector `v` such that `a @ v = w * v`. Thus, the arrays `a`, `w`, and `v` satisfy the equations `a @ v[:,i] = w[i] * v[:,i]` for `i` in `{0,...,M-1}`.

The array `v` of eigenvectors may not be of maximum rank, that is, some of the columns may be linearly dependent, although round-off error may obscure that fact. If the eigenvalues are all different, then theoretically the eigenvectors are linearly independent and `a` can be diagonalized by a similarity transformation using `v`, i.e., `inv(v) @ a @ v` is diagonal.

For non-Hermitian normal matrices the SciPy function `scipy.linalg.schur` is preferred because the matrix `v` is guaranteed to be unitary, which is not the case when using `eig`. The Schur factorization produces an upper triangular matrix rather than a diagonal matrix, but for normal matrices only the diagonal of the upper triangular matrix is needed, the rest is roundoff error.

Finally, it is emphasized that `v` consists of the *right* (as in right-hand side) eigenvectors of `a`. A vector `y` satisfying `y.T @ a = z * y.T` for some number `z` is called a *left* eigenvector of `a`, and, in general, the left and right eigenvectors of a matrix are not necessarily the (perhaps conjugate) transposes of each other.

References

-----

G. Strang, *Linear Algebra and Its Applications*, 2nd Ed., Orlando, FL, Academic Press, Inc., 1980, Various pp.

Examples

-----

```
>>> from numpy import linalg as LA
```

(Almost) trivial example with real e-values and e-vectors.

```
>>> w, v = LA.eig(np.diag((1, 2, 3)))
>>> w; v
array([1., 2., 3.])
array([[1., 0., 0.]
```

```
[0., 1., 0.],
 [0., 0., 1.]])
```

Real matrix possessing complex e-values and e-vectors; note that the e-values are complex conjugates of each other.

```
>>> w, v = LA.eig(np.array([[1, -1], [1, 1]]))
>>> w; v
array([1.+1.j, 1.-1.j])
array([[0.70710678+0.j, 0.70710678-0.j],
       [0. -0.70710678j, 0. +0.70710678j]])
```

Complex-valued matrix with real e-values (but complex-valued e-vectors); note that `a.conj().T == a`, i.e., `a` is Hermitian.

```
>>> a = np.array([[1, 1j], [-1j, 1]])
>>> w, v = LA.eig(a)
>>> w; v
array([2.+0.j, 0.+0.j])
array([[ 0. +0.70710678j, 0.70710678+0.j], # may vary
       [ 0.70710678+0.j, -0. +0.70710678j]])
```

Be careful about round-off error!

```
>>> a = np.array([[1 + 1e-9, 0], [0, 1 - 1e-9]])
>>> # Theor. e-values are 1 +/- 1e-9
>>> w, v = LA.eig(a)
>>> w; v
array([1., 1.])
array([[1., 0.],
       [0., 1.]])
```

```
In [90]: from numpy import linalg # another way to import linalg module(subpackage) from numpy
linalg.eig # now we create a name linalg to point to the linalg.py module, and can get
```

```
Out[90]: <function numpy.linalg.eig(a)>
```

```
In [91]: import numpy.linalg as LA # another way to import the linalg
LA.eig
```

```
Out[91]: <function numpy.linalg.eig(a)>
```

```
In [92]: import numpy.linalg # another way to import the linalg
numpy.linalg.eig
```

```
Out[92]: <function numpy.linalg.eig(a)>
```

```
In [93]: from numpy.linalg import eig #import the eig function directly
eig
```

```
Out[93]: <function numpy.linalg.eig(a)>
```

## Take-home message (Basic requirements)

- Understand the concept of Python modules (.py files storing objects)

- Know different ways to import modules and objects in the modules ( `import` , `import ... as` , `from ... import` )
- Understand the basic concept of package, and know how to import modules and functions within it (use `numpy` , `linalg` and `eig` as example)

## Beyond Basic Python: What's next? -- Some Suggestions

- Knowledge and wisdom
- What we have not covered in basic python: other data types (dictionary, set, tuple), input/output, exceptions, -- consult [a byte of python](#), or [programiz](#)
- The systematic book ([for example, Python Cookbook](#)) or course in computer science department (ICS-31,33)
- Practice!Practice!Practice! Useful websites such as [Leetcode](#)
- These [cheatsheets](#) from datacamp websites might also be helpful throughout this course.