

PythonCheatSheet

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1 Python Cheatsheet

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To run a cell, press **Shift+Enter** or click **Run** at the top of the page.

1.2 1. Syntax and whitespace

Python uses indented space to indicate the level of statements. The following cell is an example where 'if' and 'else' are in same level, while 'print' is separated by space to a different level. Spacing should be the same for items that are on the same level.

```
[ ]: student_number = input("Enter your student number:")
if int(student_number) != 0:
    print("Welcome student {}".format(student_number))
else:
    print("Try again!")
```

```
Enter your student number:5
Welcome student 5
```

1.3 2. Comments

In Python, comments start with hash '#' and extend to the end of the line. '#' can be at the beginning of the line or after code.

```
[ ]: # This is code to print hello world!

print("Hello world!") # Print statement for hello world
print("# is not a comment in this case")
```

Hello world!

is not a comment in this case

1.4 3. Numbers and operations

Like with other programming languages, there are four types of numbers: - Integers (e.g., 1, 20, 45, 1000) indicated by *int* - Floating point numbers (e.g., 1.25, 20.35, 1000.00) indicated by *float* - Long integers - Complex numbers (e.g., $x+2y$ where x is known)

Operation	Result
$x + y$	Sum of x and y
$x - y$	Difference of x and y
$x * y$	Product of x and y
x / y	Quotient of x and y
$x // y$	Quotient of x and y (floored)
$x \% y$	Remainder of x / y
<code>abs(x)</code>	Absolute value of x
<code>int(x)</code>	x converted to integer
<code>long(x)</code>	x converted to long integer
<code>float(x)</code>	x converted to floating point
<code>pow(x, y)</code>	x to the power y
$x ** y$	x to the power y

```
[ ]: # Number examples
a = 5 + 8
print("Sum of int numbers: {} and number format is {}".format(a, type(a)))

b = 5 + 2.3
print("Sum of int and {} and number format is {}".format(b, type(b)))
```

Sum of int numbers: 13 and number format is <class 'int'>

Sum of int and 7.3 and number format is <class 'float'>

1.5 4. String manipulation

Python has rich features like other programming languages for string manipulation.

```
[ ]: # Store strings in a variable
test_word = "hello world to everyone"

# Print the test_word value
print(test_word)

# Use [] to access the character of the string. The first character is_
↳indicated by '0'.
print(test_word[0])

# Use the len() function to find the length of the string
print(len(test_word))

# Some examples of finding in strings
print(test_word.count('l')) # Count number of times l repeats in the string
print(test_word.find("o")) # Find letter 'o' in the string. Returns the_
↳position of first match.
print(test_word.count(' ')) # Count number of spaces in the string
print(test_word.upper()) # Change the string to uppercase
print(test_word.lower()) # Change the string to lowercase
print(test_word.replace("everyone","you")) # Replace word "everyone" with "you"
print(test_word.title()) # Change string to title format
print(test_word + "!!!") # Concatenate strings
print(":".join(test_word)) # Add ":" between each character
print("".join(reversed(test_word))) # Reverse the string
```

```
hello world to everyone
h
23
3
4
3
HELLO WORLD TO EVERYONE
hello world to everyone
hello world to you
Hello World To Everyone
hello world to everyone!!!
h:e:l:l:o: :w:o:r:l:d: :t:o: :e:v:e:r:y:o:n:e
enoyreve ot dlrow olleh
```

1.6 5. Lists, tuples, and dictionaries

Python supports data types lists, tuples, dictionaries, and arrays.

1.6.1 Lists

A list is created by placing all the items (elements) inside square brackets `[]` separated by commas. A list can have any number of items, and they may be of different types (integer, float, strings, etc.).

```
[ ]: # A Python list is similar to an array. You can create an empty list too.
```

```
my_list = []  
  
first_list = [3, 5, 7, 10]  
second_list = [1, 'python', 3]
```

```
[ ]: # Nest multiple lists  
nested_list = [first_list, second_list]  
nested_list
```

```
[ ]: [[3, 5, 7, 10], [1, 'python', 3]]
```

```
[ ]: # Combine multiple lists  
combined_list = first_list + second_list  
combined_list
```

```
[ ]: [3, 5, 7, 10, 1, 'python', 3]
```

```
[ ]: # You can slice a list, just like strings  
combined_list[0:3]
```

```
[ ]: [3, 5, 7]
```

```
[ ]: # Append a new entry to the list  
combined_list.append(600)  
combined_list
```

```
[ ]: [3, 5, 7, 10, 1, 'python', 3, 600]
```

```
[ ]: # Remove the last entry from the list  
combined_list.pop()
```

```
[ ]: 600
```

```
[ ]: # Iterate the list  
for item in combined_list:  
    print(item)
```

```
3  
5  
7
```

```
10
1
python
3
```

1.6.2 Tuples

A tuple is similar to a list, but you use them with parentheses () instead of square brackets. The main difference is that a tuple is immutable, while a list is mutable.

```
[ ]: my_tuple = (1, 2, 3, 4, 5)
      my_tuple[1:4]
```

```
[ ]: (2, 3, 4)
```

1.6.3 Dictionaries

A dictionary is also known as an associative array. A dictionary consists of a collection of key-value pairs. Each key-value pair maps the key to its associated value.

```
[ ]: desk_location = {'jack': 123, 'joe': 234, 'hary': 543}
      desk_location['jack']
```

```
[ ]: 123
```

1.7 6. JSON

JSON is text written in JavaScript Object Notation. Python has a built-in package called `json` that can be used to work with JSON data.

```
[ ]: import json

      # Sample JSON data
      x = '{"first_name":"Jane", "last_name":"Doe", "age":25, "city":"Chicago"}'

      # Read JSON data
      y = json.loads(x)

      # Print the output, which is similar to a dictionary
      print("Employee name is " + y["first_name"] + " " + y["last_name"])
```

```
Employee name is Jane Doe
```

1.8 7. Loops

If, Else, Elif loop: Python supports conditional statements like any other programming language. Python relies on indentation (whitespace at the beginning of the line) to define the scope of the code.

```
[ ]: a = 22
      b = 33
      c = 100

      # if ... else example
      if a > b:
          print("a is greater than b")
      else:
          print("b is greater than a")

      # if .. else .. elif example

      if a > b:
          print("a is greater than b")
      elif b > c:
          print("b is greater than c")
      else:
          print("b is greater than a and c is greater than b")
```

b is greater than a

b is greater than a and c is greater than b

While loop: Runs a set of statements as long as the condition is true

```
[ ]: # Sample while example
      i = 1
      while i < 10:
          print("count is " + str(i))
          i += 1

      print("="*10)

      # Continue to next iteration if x is 2. Finally, print message once the
      ↪ condition is false.

      x = 0
      while x < 5:
          x += 1
          if x == 2:
              continue
          print(x)
      else:
```

```
print("x is no longer less than 5")
```

```
count is 1
count is 2
count is 3
count is 4
count is 5
count is 6
count is 7
count is 8
count is 9
```

```
=====
```

```
1
3
4
5
```

```
x is no longer less than 5
```

For loop: A For loop is more like an iterator in Python. A For loop is used for iterating over a sequence (list, tuple, dictionary, set, string, or range).

```
[ ]: # Sample for loop examples
fruits = ["orange", "banana", "apple", "grape", "cherry"]
for fruit in fruits:
    print(fruit)

print("\n")
print("="*10)
print("\n")

# Iterating range
for x in range(1, 10, 2):
    print(x)
else:
    print("task complete")

print("\n")
print("="*10)
print("\n")

# Iterating multiple lists
traffic_lights = ["red", "yellow", "green"]
action = ["stop", "slow down", "go"]

for light in traffic_lights:
    for task in action:
        print(light, task)
```

```
orange
banana
apple
grape
cherry
```

```
=====
```

```
1
3
5
7
9
task complete
```

```
=====
```

```
red stop
red slow down
red go
yellow stop
yellow slow down
yellow go
green stop
green slow down
green go
```

1.9 8. File handling

The key function for working with files in Python is the `open()` function. The `open()` function takes two parameters: filename and mode.

There are four different methods (modes) for opening a file:

- "r" - Read
- "a" - Append
- "w" - Write
- "x" - Create

In addition, you can specify if the file should be handled in binary or text mode.

- "t" - Text
- "b" - Binary


```
[ ]: # Let's create a test text file
!echo "This is a test file with text in it. This is the first line." > test.txt
!echo "This is the second line." >> test.txt
!echo "This is the third line." >> test.txt
```

```
[ ]: # Read file
file = open('test.txt', 'r')
print(file.read())
file.close()

print("\n")
print("="*10)
print("\n")

# Read first 10 characters of the file
file = open('test.txt', 'r')
print(file.read(10))
file.close()

print("\n")
print("="*10)
print("\n")

# Read line from the file

file = open('test.txt', 'r')
print(file.readline())
file.close()
```

This is a test file with text in it. This is the first line.
This is the second line.
This is the third line.

=====

This is a

=====

This is a test file with text in it. This is the first line.

```
[ ]: # Create new file

file = open('test2.txt', 'w')
file.write("This is content in the new test2 file.")
file.close()

# Read the content of the new file
file = open('test2.txt', 'r')
print(file.read())
file.close()
```

This is content in the new test2 file.

```
[ ]: # Update file
file = open('test2.txt', 'a')
file.write("\nThis is additional content in the new file.")
file.close()

# Read the content of the new file
file = open('test2.txt', 'r')
print(file.read())
file.close()
```

This is content in the new test2 file.

This is additional content in the new file.

```
[ ]: # Delete file
import os
file_names = ["test.txt", "test2.txt"]
for item in file_names:
    if os.path.exists(item):
        os.remove(item)
        print(f"File {item} removed successfully!")
    else:
        print(f"{item} file does not exist.")
```

File test.txt removed successfully!

File test2.txt removed successfully!

1.10 9. Functions

A function is a block of code that runs when it is called. You can pass data, or *parameters*, into the function. In Python, a function is defined by `def`.

```
[ ]: # Defining a function
def new_func():
    print("A simple function")
```

```
# Calling the function
new_func()
```

A simple function

```
[ ]: # Sample fuction with parameters

def param_func(first_name):
    print(f"Employee name is {first_name}.")

param_func("Harry")
param_func("Larry")
param_func("Shally")
```

Employee name is Harry.
Employee name is Larry.
Employee name is Shally.

Anonymous functions (lambda): A lambda is a small anonymous function. A lambda function can take any number of arguments but only one expression.

```
[ ]: # Sample lambda example
x = lambda y: y + 100
print(x(15))

print("\n")
print("="*10)
print("\n")

x = lambda a, b: a*b/100
print(x(2,4))
```

115

=====

0.08

1.11 10. Working with datetime

A datetime module in Python can be used to work with date objects.

```
[ ]: import datetime

x = datetime.datetime.now()
```

```
print(x)
print(x.year)
print(x.strftime("%A"))
print(x.strftime("%B"))
print(x.strftime("%d"))
print(x.strftime("%H:%M:%S %p"))
```

```
2022-03-10 20:18:59.907021
2022
Thursday
March
10
20:18:59 PM
```

1.12 11. NumPy

NumPy is the fundamental package for scientific computing with Python. Among other things, it contains:

- Powerful N-dimensional array object
- Sophisticated (broadcasting) functions
- Tools for integrating C/C++ and Fortran code
- Useful linear algebra, Fourier transform, and random number capabilities

```
[ ]: # Install NumPy using pip
!pip install --upgrade pip
!pip install numpy
```

```
Requirement already satisfied: pip in /usr/local/lib/python3.7/dist-packages
(22.0.4)
```

```
WARNING: Running pip as the 'root' user can result in broken permissions
and conflicting behaviour with the system package manager. It is recommended to
use a virtual environment instead: https://pip.pypa.io/warnings/venv
```

```
Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-
packages (1.21.5)
```

```
WARNING: Running pip as the 'root' user can result in broken permissions
and conflicting behaviour with the system package manager. It is recommended to
use a virtual environment instead: https://pip.pypa.io/warnings/venv
```

```
[ ]: # Import NumPy module
import numpy as np
```

1.12.1 Inspecting your array

```
[ ]: # Create array
a = np.arange(15).reshape(3, 5) # Create array with range 0-14 in 3 by 5
    ↪ dimension
b = np.zeros((3,5)) # Create array with zeroes
c = np.ones( (2,3,4), dtype=np.int16 ) # Create array with ones and defining
    ↪ data types
d = np.ones((3,5))
```

```
[ ]: a.shape # Array dimension
```

```
[ ]: (3, 5)
```

```
[ ]: len(b) # Length of array
```

```
[ ]: 3
```

```
[ ]: c.ndim # Number of array dimensions
```

```
[ ]: 3
```

```
[ ]: a.size # Number of array elements
```

```
[ ]: 15
```

```
[ ]: b.dtype # Data type of array elements
```

```
[ ]: dtype('float64')
```

```
[ ]: c.dtype.name # Name of data type
```

```
[ ]: 'int16'
```

```
[ ]: c.astype(float) # Convert an array type to a different type
```

```
[ ]: array([[1., 1., 1., 1.],
           [1., 1., 1., 1.],
           [1., 1., 1., 1.]],

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

1.12.2 Basic math operations

```
[ ]: # Create array
a = np.arange(15).reshape(3, 5) # Create array with range 0-14 in 3 by 5
    ↳dimension
b = np.zeros((3,5)) # Create array with zeroes
c = np.ones( (2,3,4), dtype=np.int16 ) # Create array with ones and defining
    ↳data types
d = np.ones((3,5))
```

```
[ ]: np.add(a,b) # Addition
```

```
[ ]: array([[ 0.,  1.,  2.,  3.,  4.],
           [ 5.,  6.,  7.,  8.,  9.],
           [10., 11., 12., 13., 14.]])
```

```
[ ]: np.subtract(a,b) # Subtraction
```

```
[ ]: array([[ 0.,  1.,  2.,  3.,  4.],
           [ 5.,  6.,  7.,  8.,  9.],
           [10., 11., 12., 13., 14.]])
```

```
[ ]: np.divide(a,d) # Division
```

```
[ ]: array([[ 0.,  1.,  2.,  3.,  4.],
           [ 5.,  6.,  7.,  8.,  9.],
           [10., 11., 12., 13., 14.]])
```

```
[ ]: np.multiply(a,d) # Multiplication
```

```
[ ]: array([[ 0.,  1.,  2.,  3.,  4.],
           [ 5.,  6.,  7.,  8.,  9.],
           [10., 11., 12., 13., 14.]])
```

```
[ ]: np.array_equal(a,b) # Comparison - arraywise
```

```
[ ]: False
```

1.12.3 Aggregate functions

```
[ ]: # Create array
a = np.arange(15).reshape(3, 5) # Create array with range 0-14 in 3 by 5
    ↳dimension
b = np.zeros((3,5)) # Create array with zeroes
c = np.ones( (2,3,4), dtype=np.int16 ) # Create array with ones and defining
    ↳data types
```

```
d = np.ones((3,5))
```

```
[ ]: a.sum() # Array-wise sum
```

```
[ ]: 105
```

```
[ ]: a.min() # Array-wise min value
```

```
[ ]: 0
```

```
[ ]: a.mean() # Array-wise mean
```

```
[ ]: 7.0
```

```
[ ]: a.max(axis=0) # Max value of array row
```

```
[ ]: array([10, 11, 12, 13, 14])
```

```
[ ]: np.std(a) # Standard deviation
```

```
[ ]: 4.320493798938574
```

1.12.4 Subsetting, slicing, and indexing

```
[ ]: # Create array
a = np.arange(15).reshape(3, 5) # Create array with range 0-14 in 3 by 5
    ↳ dimension
b = np.zeros((3,5)) # Create array with zeroes
c = np.ones( (2,3,4), dtype=np.int16 ) # Create array with ones and defining
    ↳ data types
d = np.ones((3,5))
```

```
[ ]: a[1,2] # Select element of row 1 and column 2
```

```
[ ]: 7
```

```
[ ]: a[0:2] # Select items on index 0 and 1
```

```
[ ]: array([[0, 1, 2, 3, 4],
          [5, 6, 7, 8, 9]])
```

```
[ ]: a[:1] # Select all items at row 0
```

```
[ ]: array([[0, 1, 2, 3, 4]])
```

```
[ ]: a[-1:] # Select all items from last row
```

```
[ ]: array([[10, 11, 12, 13, 14]])
```

```
[ ]: a[a<2] # Select elements from 'a' that are less than 2
```

```
[ ]: array([0, 1])
```

1.12.5 Array manipulation

```
[ ]: # Create array
a = np.arange(15).reshape(3, 5) # Create array with range 0-14 in 3 by 5
    ↪ dimension
b = np.zeros((3,5)) # Create array with zeroes
c = np.ones( (2,3,4), dtype=np.int16 ) # Create array with ones and defining
    ↪ data types
d = np.ones((3,5))
```

```
[ ]: np.transpose(a) # Transpose array 'a'
```

```
[ ]: array([[ 0,  5, 10],
           [ 1,  6, 11],
           [ 2,  7, 12],
           [ 3,  8, 13],
           [ 4,  9, 14]])
```

```
[ ]: a.ravel() # Flatten the array
```

```
[ ]: array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14])
```

```
[ ]: a.reshape(5,-2) # Reshape but don't change the data
```

```
[ ]: array([[ 0,  1,  2],
           [ 3,  4,  5],
           [ 6,  7,  8],
           [ 9, 10, 11],
           [12, 13, 14]])
```

```
[ ]: np.append(a,b) # Append items to the array
```

```
[ ]: array([ 0.,  1.,  2.,  3.,  4.,  5.,  6.,  7.,  8.,  9., 10., 11., 12.,
           13., 14.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
           0.,  0.,  0.,  0.])
```

```
[ ]: np.concatenate((a,d), axis=0) # Concatenate arrays
```

```
[ ]: array([[ 0.,  1.,  2.,  3.,  4.],
           [ 5.,  6.,  7.,  8.,  9.]])
```



```
[10., 11., 12., 13., 14.],
 [ 1.,  1.,  1.,  1.,  1.],
 [ 1.,  1.,  1.,  1.,  1.],
 [ 1.,  1.,  1.,  1.,  1.]])
```

```
[ ]: np.vsplit(a,3) # Split array vertically at 3rd index
```

```
[ ]: [array([[0, 1, 2, 3, 4]]),
      array([[5, 6, 7, 8, 9]]),
      array([[10, 11, 12, 13, 14]])]
```

```
[ ]: np.hsplit(a,5) # Split array horizontally at 5th index
```

```
[ ]: [array([[ 0],
             [ 5],
             [10]]), array([[ 1],
                             [ 6],
                             [11]]), array([[ 2],
                                             [ 7],
                                             [12]]), array([[ 3],
                                                             [ 8],
                                                             [13]]), array([[ 4],
                                                         [ 9],
                                                         [14]])]
```

1.13 Pandas

Pandas is an open source, BSD-licensed library providing high-performance, easy-to-use data structures and data analysis tools for the Python programming language.

Pandas DataFrames are the most widely used in-memory representation of complex data collections within Python.

```
[ ]: # Install pandas, xlrd, and openpyxl using pip
!pip install pandas
!pip install xlrd openpyxl
```

```
Requirement already satisfied: pandas in /usr/local/lib/python3.7/dist-packages
(1.3.5)
```

```
Requirement already satisfied: pytz>=2017.3 in /usr/local/lib/python3.7/dist-
packages (from pandas) (2018.9)
```

```
Requirement already satisfied: python-dateutil>=2.7.3 in
/usr/local/lib/python3.7/dist-packages (from pandas) (2.8.2)
```

```
Requirement already satisfied: numpy>=1.17.3 in /usr/local/lib/python3.7/dist-
packages (from pandas) (1.21.5)
```

```
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.7/dist-
packages (from python-dateutil>=2.7.3->pandas) (1.15.0)
```

WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommended to use a virtual environment instead: <https://pip.pypa.io/warnings/venv>

Requirement already satisfied: xlrd in /usr/local/lib/python3.7/dist-packages (1.1.0)

Requirement already satisfied: openpyxl in /usr/local/lib/python3.7/dist-packages (3.0.9)

Requirement already satisfied: et-xmlfile in /usr/local/lib/python3.7/dist-packages (from openpyxl) (1.1.0)

WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommended to use a virtual environment instead: <https://pip.pypa.io/warnings/venv>

```
[ ]: # Import NumPy and Pandas modules
import numpy as np
import pandas as pd
```

```
[ ]: # Sample dataframe df
df = pd.DataFrame({'num_legs': [2, 4, np.nan, 0],
                   'num_wings': [2, 0, 0, 0],
                   'num_specimen_seen': [10, np.nan, 1, 8]},
                  index=['falcon', 'dog', 'spider', 'fish'])
df # Display dataframe df
```

```
[ ]:
```

	num_legs	num_wings	num_specimen_seen
falcon	2.0	2	10.0
dog	4.0	0	NaN
spider	NaN	0	1.0
fish	0.0	0	8.0

```
[ ]: # Another sample dataframe df1 - using NumPy array with datetime index and
↳ labeled column
df1 = pd.date_range('20130101', periods=6)
df1 = pd.DataFrame(np.random.randn(6, 4), index=df1, columns=list('ABCD'))
df1 # Display dataframe df1
```

```
[ ]:
```

	A	B	C	D
2013-01-01	0.585722	-0.649741	-3.578389	1.256532
2013-01-02	0.365194	-0.030423	-1.903047	-1.967513
2013-01-03	-1.273320	-0.608034	-0.118732	-0.044090
2013-01-04	-1.352582	1.306480	0.064463	-0.156061
2013-01-05	1.675461	-0.836540	-0.209184	0.012749
2013-01-06	0.006478	-0.977767	1.213337	-0.512744

1.13.1 Viewing data

```
[ ]: df1 = pd.date_range('20130101', periods=6)
df1 = pd.DataFrame(np.random.randn(6, 4), index=df1, columns=list('ABCD'))
```

```
[ ]: df1.head(2) # View top data
```

```
[ ]:
```

	A	B	C	D
2013-01-01	0.856162	-0.967504	-1.239942	0.992299
2013-01-02	2.456804	1.249084	0.580066	-0.827189

```
[ ]: df1.tail(2) # View bottom data
```

```
[ ]:
```

	A	B	C	D
2013-01-05	0.146800	-0.728447	1.299218	1.283088
2013-01-06	-0.358769	-0.986644	-0.193084	-0.282883

```
[ ]: df1.index # Display index column
```

```
[ ]: DatetimeIndex(['2013-01-01', '2013-01-02', '2013-01-03', '2013-01-04',
                    '2013-01-05', '2013-01-06'],
                    dtype='datetime64[ns]', freq='D')
```

```
[ ]: df1.dtypes # Inspect datatypes
```

```
[ ]: A    float64
B    float64
C    float64
D    float64
dtype: object
```

```
[ ]: df1.describe() # Display quick statistics summary of data
```

```
[ ]:
```

	A	B	C	D
count	6.000000	6.000000	6.000000	6.000000
mean	0.926104	-0.568749	-0.038979	0.496867
std	0.979895	0.936510	0.899033	0.846609
min	-0.358769	-1.418179	-1.239942	-0.827189
25%	0.324140	-0.981859	-0.552038	-0.010814
50%	1.002041	-0.847976	-0.100762	0.898845
75%	1.267759	-0.602716	0.432939	1.005947
max	2.456804	1.249084	1.299218	1.283088

1.13.2 Subsetting, slicing, and indexing

```
[ ]: df1 = pd.date_range('20130101', periods=6)
df1 = pd.DataFrame(np.random.randn(6, 4), index=df1, columns=list('ABCD'))
```

```
[ ]: df1.T # Transpose data
```

```
[ ]:      2013-01-01  2013-01-02  2013-01-03  2013-01-04  2013-01-05  2013-01-06
A      0.624376   -0.032064   -0.477969    2.134176    0.171814    0.060342
B     -0.688370   -0.467953    0.283626    0.980079   -0.335895    0.889149
C      0.246922   -1.080885    0.205450   -1.205140   -0.919308    0.602191
D      3.277770   -1.113337    1.864838    0.487757   -1.183843   -0.208107
```

```
[ ]: df1.sort_index(axis=1, ascending=False) # Sort by an axis
```

```
[ ]:      D      C      B      A
2013-01-01  3.277770  0.246922 -0.688370  0.624376
2013-01-02 -1.113337 -1.080885 -0.467953 -0.032064
2013-01-03  1.864838  0.205450  0.283626 -0.477969
2013-01-04  0.487757 -1.205140  0.980079  2.134176
2013-01-05 -1.183843 -0.919308 -0.335895  0.171814
2013-01-06 -0.208107  0.602191  0.889149  0.060342
```

```
[ ]: df1.sort_values(by='B') # Sort by values
```

```
[ ]:      A      B      C      D
2013-01-01  0.624376 -0.688370  0.246922  3.277770
2013-01-02 -0.032064 -0.467953 -1.080885 -1.113337
2013-01-05  0.171814 -0.335895 -0.919308 -1.183843
2013-01-03 -0.477969  0.283626  0.205450  1.864838
2013-01-06  0.060342  0.889149  0.602191 -0.208107
2013-01-04  2.134176  0.980079 -1.205140  0.487757
```

```
[ ]: df1['A'] # Select column A
```

```
[ ]: 2013-01-01    0.624376
2013-01-02   -0.032064
2013-01-03   -0.477969
2013-01-04    2.134176
2013-01-05    0.171814
2013-01-06    0.060342
Freq: D, Name: A, dtype: float64
```

```
[ ]: df1[0:3] # Select index 0 to 2
```

```
[ ]:      A      B      C      D
2013-01-01  0.624376 -0.688370  0.246922  3.277770
```

```
2013-01-02 -0.032064 -0.467953 -1.080885 -1.113337
2013-01-03 -0.477969  0.283626  0.205450  1.864838
```

```
[ ]: df1['20130102':'20130104'] # Select from index matching the values
```

```
[ ]:
      A      B      C      D
2013-01-02 -0.032064 -0.467953 -1.080885 -1.113337
2013-01-03 -0.477969  0.283626  0.205450  1.864838
2013-01-04  2.134176  0.980079 -1.205140  0.487757
```

```
[ ]: df1.loc[:, ['A', 'B']] # Select on a multi-axis by label
```

```
[ ]:
      A      B
2013-01-01  0.624376 -0.688370
2013-01-02 -0.032064 -0.467953
2013-01-03 -0.477969  0.283626
2013-01-04  2.134176  0.980079
2013-01-05  0.171814 -0.335895
2013-01-06  0.060342  0.889149
```

```
[ ]: df1.iloc[3] # Select via the position of the passed integers
```

```
[ ]: A    2.134176
      B    0.980079
      C   -1.205140
      D    0.487757
      Name: 2013-01-04 00:00:00, dtype: float64
```

```
[ ]: df1[df1 > 0] # Select values from a DataFrame where a boolean condition is met
```

```
[ ]:
      A      B      C      D
2013-01-01  0.624376    NaN  0.246922  3.277770
2013-01-02    NaN    NaN    NaN    NaN
2013-01-03    NaN  0.283626  0.205450  1.864838
2013-01-04  2.134176  0.980079    NaN  0.487757
2013-01-05  0.171814    NaN    NaN    NaN
2013-01-06  0.060342  0.889149  0.602191    NaN
```

```
[ ]: df2 = df1.copy() # Copy the df1 dataset to df2
df2['E'] = ['one', 'one', 'two', 'three', 'four', 'three'] # Add column E with
↳ value
df2[df2['E'].isin(['two', 'four'])] # Use isin method for filtering
```

```
[ ]:
      A      B      C      D      E
2013-01-03 -0.477969  0.283626  0.205450  1.864838  two
2013-01-05  0.171814 -0.335895 -0.919308 -1.183843  four
```

1.13.3 Missing data

Pandas primarily uses the value `np.nan` to represent missing data. It is not included in computations by default.

```
[ ]: df = pd.DataFrame({'num_legs': [2, 4, np.nan, 0],  
                        'num_wings': [2, 0, 0, 0],  
                        'num_specimen_seen': [10, np.nan, 1, 8]},  
                        index=['falcon', 'dog', 'spider', 'fish'])
```

```
[ ]: df.dropna(how='any') # Drop any rows that have missing data
```

```
[ ]:      num_legs  num_wings  num_specimen_seen  
falcon      2.0          2          10.0  
fish        0.0          0           8.0
```

```
[ ]: df.dropna(how='any', axis=1) # Drop any columns that have missing data
```

```
[ ]:      num_wings  
falcon          2  
dog             0  
spider          0  
fish            0
```

```
[ ]: df.fillna(value=5) # Fill missing data with value 5
```

```
[ ]:      num_legs  num_wings  num_specimen_seen  
falcon      2.0          2          10.0  
dog         4.0          0           5.0  
spider      5.0          0           1.0  
fish        0.0          0           8.0
```

```
[ ]: pd.isna(df) # To get boolean mask where data is missing
```

```
[ ]:      num_legs  num_wings  num_specimen_seen  
falcon      False      False          False  
dog         False      False          True  
spider      True       False          False  
fish        False      False          False
```

1.13.4 File handling

```
[ ]: df = pd.DataFrame({'num_legs': [2, 4, np.nan, 0],  
                        'num_wings': [2, 0, 0, 0],  
                        'num_specimen_seen': [10, np.nan, 1, 8]},  
                        index=['falcon', 'dog', 'spider', 'fish'])
```

```
[ ]: df.to_csv('foo.csv') # Write to CSV file
```

```
[ ]: pd.read_csv('foo.csv') # Read from CSV file
```

```
[ ]: Unnamed: 0  num_legs  num_wings  num_specimen_seen
0      falcon      2.0      2          10.0
1         dog      4.0      0           NaN
2      spider      NaN      0           1.0
3        fish      0.0      0           8.0
```

```
[ ]: df.to_excel('foo.xlsx', sheet_name='Sheet1') # Write to Microsoft Excel file
```

```
[ ]: pd.read_excel('foo.xlsx', 'Sheet1', index_col=None, na_values=['NA'],
↪engine='openpyxl') # Read from Microsoft Excel file
```

```
[ ]: Unnamed: 0  num_legs  num_wings  num_specimen_seen
0      falcon      2.0      2          10.0
1         dog      4.0      0           NaN
2      spider      NaN      0           1.0
3        fish      0.0      0           8.0
```

1.13.5 Plotting

```
[ ]: # Install Matplotlib using pip
!pip install matplotlib
```

Requirement already satisfied: matplotlib in /usr/local/lib/python3.7/dist-packages (3.2.2)

Requirement already satisfied: cyclor>=0.10 in /usr/local/lib/python3.7/dist-packages (from matplotlib) (0.11.0)

Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /usr/local/lib/python3.7/dist-packages (from matplotlib) (3.0.7)

Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.7/dist-packages (from matplotlib) (1.3.2)

Requirement already satisfied: python-dateutil>=2.1 in /usr/local/lib/python3.7/dist-packages (from matplotlib) (2.8.2)

Requirement already satisfied: numpy>=1.11 in /usr/local/lib/python3.7/dist-packages (from matplotlib) (1.21.5)

Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.7/dist-packages (from python-dateutil>=2.1->matplotlib) (1.15.0)

WARNING: Running pip as the 'root' user can result in broken permissions

and conflicting behaviour with the system package manager. It is recommended to

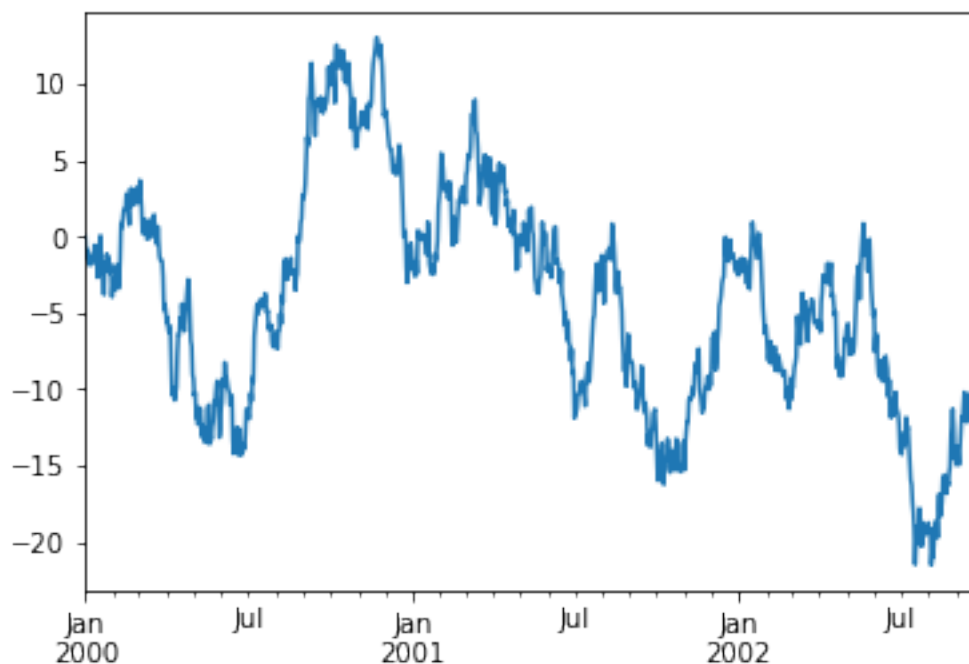
use a virtual environment instead: <https://pip.pypa.io/warnings/venv>

```
[ ]: from matplotlib import pyplot as plt # Import Matplotlib module
```

```
[ ]: # Generate random time-series data
ts = pd.Series(np.random.randn(1000), index=pd.date_range('1/1/2000',
    ↳ periods=1000))
ts.head()
```

```
[ ]: 2000-01-01    -0.029855
     2000-01-02    -1.234720
     2000-01-03     0.406091
     2000-01-04    -0.719265
     2000-01-05     0.443042
     Freq: D, dtype: float64
```

```
[ ]: ts = ts.cumsum()
     ts.plot() # Plot graph
     plt.show()
```



```
[ ]: # On a DataFrame, the plot() method is convenient to plot all of the columns
    ↳ with labels
df4 = pd.DataFrame(np.random.randn(1000, 4), index=ts.index, columns=['A', 'B',
    ↳ 'C', 'D'])
df4 = df4.cumsum()
df4.head()
```



```
[ ]:
```

	A	B	C	D
2000-01-01	-1.750676	-0.631666	-0.286245	0.116738
2000-01-02	-0.945290	-2.954609	-0.315168	0.882456
2000-01-03	-1.763598	-2.834895	0.001486	0.118155
2000-01-04	-3.152844	-4.084474	1.848818	1.086287
2000-01-05	-2.210792	-4.873050	1.552365	1.971197

```
[ ]: df4.plot()  
plt.show()
```

