

*) shell commands can be executed in google colab. using the '!' as prefix in front of the command.

*) to know the information about the cpu assigned to you on colab.

↓
we look in the proc folder, at the file cpuinfo.

↓
similarly for memory info → meminfo inside proc folder

*) sections can be used to write neat & tidy code in colab.

*) Latex can also be used in the markup version of a cell in colab for better documentation.

*) Adding '?' as suffix to any function will open colab help. and will give required information on how to operate the function.

help(fn-name)

*) Inside functions, the text inside the triple quotes is used as the docstring & is used as reference when '?' is used with that function.

↑
↓
Hence a good practice to write or provide a suitable docstring with each function.

Colab pointers

Python pointers

*) just assigning a list to another variable leads to a shallow copy.

obj1 = list(obj)

obj2 = obj.copy

↓
deep copies

list → []

tuple → ()

set, dict → {}

↓
better to declare explicitly with name().
↓
defaults to dict.

*) + operator concatenates a list to another

*) list change (s)



~~list~~ returns a list of elements
from 0, 4 (both inclusive)

*) map function → returns a list of the results after applying the given fn. to each element of the given iterable object.



`b = list (map (lambda x : x**2, a))`

important to typecast



special kind of
function that takes
one argument and
operates on it



more like a one line
function in python

parameter → defn.

arguments → actual sent
data.

*) filter function → applies fn. sent as parameter on each element of the list, and returns the ones which return true.



`c = list (filter (lambda x : x%2 == 0, b))`

*) tuple : immutable as compared to list



supposed to be faster than a list for computations.

*) time package can be employed for time based experiments.



`time.time()` → returns current time in
seconds since epoch

*) set → unique entities. (sorted as well)

↓
faster queries

↓
(element in iterable) → syntax to check
for element in object

↓
set.add ()

returns bool.

↓
to add new elements in set.

*) my-dict.keys(), my-dict.values()

↓

to access keys and
values in a dictionary
↓
respective type of object
for each
Can be typecasted
to a list

*) my-dict.items()

↓

to access key, value pairs in a dictionary

*) python packages → import math as m
from math import factorial as fact

↓
this can be any .py file, and any
function or variable / object can be
imported.

*) File handling in python :

file = open('mobile-cleaned.csv', 'r')

↓
file name

↓
opening mode

↓
can be 'r' or 'w'

s = file.readline()

↓
reads one line in the file

*) `string.split(delimiter)`

↓
splits the given string
based on the delimiter
and returns a list
of separated elements

*) `file.close()`

↓
lines can't be read after this
is executed, until file is
opened again

*) `with open('mobile-cleaned.csv', 'r') as file:`
`print(file.readline())`

↓
the file is scoped
and not accessible
outside this scope

*) `with open('mobile-cleaned.csv', 'r') as file:`
`print(file.read())`

↓
reads all lines.

for line in file:

`print(line)` → prints each line
separately

*) `with open('my-first-file-output.txt', 'w') as file:`
`file.write('hello world from py code')`

*) class → keyword to create classes

•) github notes on oops.

•) class MobilePhone():

def __init__(self, name):

self.name = name

→ instance of the class

↓
analogous to this

•) MobilePhone.__doc__

inherits all fns.

↓
to access doc string of the class



•) inheritance → class iPhone(MobilePhone):

def __init__(self, name):

MobilePhone.__init__(self, name,
false, 4)

explicit
overwriting
required.

MRO: Method Resolution

order

(ops left to right)

↓
calling superclasses'
init method

•) __str__(self):

return str.object → accessed or used
when print(obj)
is done

*) Numpy → import numpy as np. → alias.

x = np.array(list)

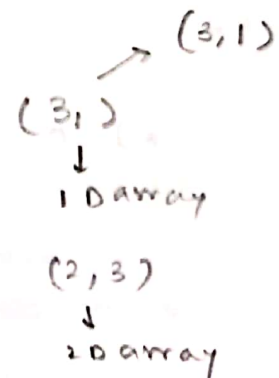
type(x) → numpy.ndarray

x = np.asarray(1, float)

x.shape()

x[0,0] → accessing elements in np array

x[0,0:2] → accessing using ranges.



*) inbuilt arrays

↓
 $x = \text{np.zeros}((4, 5))$

↑
tuple entry

$x = \text{np.eye}(4, 5)$

↓
diagonal entries are 1,
others marked with 0

$x = \text{np.random.random}((4, 5))$

↓
random no. elements
between 0 & 1

↓
apt. addition / subtraction
and multiplication can be
done for changing & altering
range of random no.s

*) $z = x.T$ (taking transpose of x)

*) reshaping np arrays

↓
 $y = x.reshape(20, 1)$

*) $A = \text{np.arange}(5)$

↓
analogous to array
from 0 to 4. (both
inclusive)

* > $A = \text{np.arange}(5)$

$B = \text{np.arange}(5)$

$C = A + B$

$D = A - B$

$E = A * B$

→ point / element wise operation.

* > $A + 1 \rightarrow$ scalar (1) gets broadcasted about the vector / np.array.

* > $A = \text{np.floor}(\text{np.random.random}((2, 3)) * 10)$

↓
generating a random matrix of 2×3 with elements b/w 0 & 1.

↓
multiplied by 10

↓
taking floor of those values.

↓
integers between 0 & 10
(0 inclusive, 10 not inclusive)

* > $A.\text{sum}()$ → sum all elements of an array

$A.\text{sum}(\text{axis}=0)$ → sum elements row wise

$A.\text{sum}(\text{axis}=1)$

↳ sum elements column wise

* > axes are governed by dimensions in the array.

axis = 0 (row)

axis = 1 (cols)

axis = 2 (obj. point to point)

axis = 3 (obj to obj point to point)

} reference objects change.

* `np.inner()` $(5, 4) \cdot (3, 4)$
 \downarrow
 $(5, 3)$
 product takes / picks elements
 of one array and multiplies
 with opp. ended row & then
 takes sum.

* `np.dot()` \rightarrow conventional dot product

* `np.outer()` \rightarrow without taking sum,
 each element for itself

$$(5, 4) \cdot (3, 4)$$

$$(20, 12)$$

* Plotting (Matplotlib)

• `import matplotlib.pyplot as plt` \rightarrow to use

• `plt.plot(x)`

`plt.show()`

matplotlib lib
 \downarrow
 package needs to
 be imported.

if x is a list or

`np.array`, then the

values act as y axis values

and indices act as x axis
 values.

\downarrow
 these values are plotted
 and then `show` is used to
 print / show the graph

• `plt.plot(x, '*')`

\downarrow
 scatter/plots x with the corresponding
 sign/symbol.

`plt.plot(x, 'x-')`

\downarrow
 both line and symbol get plotted

•) `x = np.linspace(0, 10, 100)`

↓ ↓ ↓
range no. of elements
in the given range
returns a numpy
array which returns
given no. of elements
in the given range.

! pip install
seaborn == 0.9.0
installing packages
with specific
versions

`y = np.power(x, 0.5)`

`plt.plot(x, y)` → plots `x` vs `y` as per the given
`plt.show()` numpy array, which should
indicate the sqrt distribution

→ use of an extension on top of matplotlib to make the
plots look better
↓
seaborn.

•) import seaborn as sns → wraps around matplotlib
and makes the plots look
more presentable.
•) `sns.set()` → sets up properties
of matplotlib for better plotting.

•) `sns.lineplot(x, y)` sns -- version --
`plt.show()`

→ better data framing package

*) pandas : import pandas as pd

`data = pd.read_csv('mobile-cleaned.csv')`
`data.head()`

•) `ax` → used as handles for axes.

`ax = sns.scatterplot(x="standby-time", y="battery-capacity", data=data)`

! better to have more dense plots,

hue can be used to get another
feature as part of the plot

`ax = sns.scatterplot(x="stand-by-time", y="battery-capacity",
hue="thickness", data=data)`

•) histogram using sns. → distplot

```
ax = sns.distplot (data ['stand-by-time'])
```



bars + distribution

kde = False → no distribution

rug = True

↳ will show where exactly are the values.

bins = 20

↳ can be used to manipulate the no. of bins in a graph

columns with specific values



•) boxplot

→ all possible values with a common plane for common range of values

```
ax = sns.boxplot (x = 'is-liked',
```

```
y = 'battery_capacity',
```

```
data = data)
```

```
ax = sns.boxplot (x = 'expandable-memory',
```

```
y = 'price', data = data)
```



can be used to check the relation b/w two cols.

only 2d o/p.



•) heatmap

uniform-data =

```
np.random.random ((10,12))
```

```
ax = sns.heatmap (uniform-data)
```



values in 2D matrix mapped to a color map



good method to visualize the performance of parameters

→ changing color of heatmap

ax = sns.heatmap(uniform_data, cmap="yerglb")

* plotting images.

specific subsection of
matplotlib to handle
images
import matplotlib.image as mpimg

i) reading image

img = mpimg.imread('one-fourth-labs.png')

matrix → color patterns at different pixels.

ii) imgplot = plt.imshow(img)

↳ plotting image