

```
#1 "if __name__ == '__main__':"
```

- This line is typically used to check whether the python script is being run directly or being imported as a module into another script. When a python file is run directly, the special variable '\_\_name\_\_' is set to '\_\_main\_\_'

Example, how it is commonly used

```
def main():
    check
    print("Hi")
if __name__ == '__main__':
    main()
```

Example, using class

```
class MyClass:
    def __init__(self, name):
        self.name = name
    def greet(self):
        print("Hello, " + self.name)
def main():
    obj = MyClass("Shiva")
    obj.greet()
if __name__ == '__main__':
    main()
```

#2

c = map(float, input().split())

Syntax: map(function-name, sequence)

- This line of code is used to read multiple input values from the user in a single line, split them into separate strings based on whitespace, convert each string to a float and stores these float values in an iterable object ('map' object)

- input() - Reads a line of input <sup>from</sup> the user as string
- split() - splits the input string into a list of substrings based on whitespace
- float() - converts string datatype to float datatype
- map(float, -) - applies the float() function to each element produced by 'split()'

Example :

```
c = map(float, input("Enter : ").split())
Print(type(c))
for i in c:
    Print(i)
```

Output :

Enter : 1.1 1.3 2.5

<class 'map'>

1.1

1.3

2.5

@#5

#3

## filter function using Lambda function

- inbuilt function 'filter' filters the elements of the sequence for which the function returns true. it creates a list of the elements for which the function evaluates to true and ignores the other elements.

Syntax :

filter (function\_name, sequence)

function\_name → name of the function or entire lambda function  
 sequence → list of elements

Example with normal user-defined function, (to filter odd numbers)def odd ~~xxxx~~(x):

```
if x%2 == 0:
    return False False
else:
    return True True
```

Output:

Enter: 1 2 6 7 3 22 11

[1, 7, 3, 11]

```
Elements = [int(x) for x in input("Enter:").split()]
result = list(filter(odd xxxx, Elements))
```

Print(result)

Example with lambda function (filters even)

```
Elements = [int(x) for x in input("Enter:").split()]
```

```
result = list(filter(lambda x: (x%2 == 0), Elements))
```

Print(result)

Output:

Enter: 1 2 6 7 3 22 11

~~[~~ 2, 6, 22 ]

#4

`lst = list(int(x) for x in input().split())`

`input()` → This function reads a line of input from the user.

`.split()` → This function splits the input into a list of substrings based on the whitespace by default (separates with comma)

`int(x) for x in input().split()` → This is a generator expression that iterates over the list of substrings produced by `split()`, converting each substring ~~as~~  $x$  to an integer.

`list(...)` → This function takes the generator expression and constructs a list from its values

Example :

`lst = list(int(x) for x in input("Enter :").split())`

`print(lst)`

Output :

Enter : 1 3 6 7 8 11 12  
[1, 3, 6, 7, 8, 11, 12]

## #5 Map function using ~~lambda~~

- inbuilt function 'map' creates a list of modified elements returned by the function. The map function passes the ~~the~~ individual element from the sequence to the function and creates a sequence of the returned elements (modified elements)

Syntax:

`map(function_name, sequence)`

function\_name → name of the function or entire lambda function  
sequence → list of elements

Example with user defined function, (squaring the elements)

```
def power(x):  
    return x**2
```

```
lst = [int(x) for x in input("Enter:").split()]
```

```
k = list(map(power, lst))
```

```
print('updated:', k)
```

Output:

Enter: 1 2 3 4

updated: [1, 4, 9, 16]

Example with Lambda

```
lst = [int(x) for x in input("Enter:").split()]
```

```
k = list(map(lambda x: (x**2), lst))
```

```
print('updated:', k)
```

Output:

Enter: 1 2 3 4 5

updated: [1, 4, 9, 16, 25]

#6

## Reduce function

- inbuilt function that reduces the list of elements to a single value. The reduction is done by according to the processing performed by the function. Reduce takes two values from the sequence passed to it in the first step and then passes two values to the function to reduce and then it passes the returned result and one value from the list to the function to reduce it to a single value again. This process is continued until the last element of the list not processed. The map function is used to pass two individual values from the sequence passed to it. reduce function is defined in the functools module syntax; we need to import functools to use reduce function  
 $\text{reduce}(\text{function\_name}, \text{sequence})$

Example using user-defined function,

from functools import \*

def add(x, y):

    return x+y

lst = [int(x) for x in input("Enter:").split()]

k = reduce(add, lst)

print('Result:', k)

Output:

Enter: 1 2 3 4 5  
Result: 15

Example using Lambda,

from functools import \*

lst = [int(x) for x in input("Enter:").split()]

k = reduce(lambda x, y: (x+y), lst)

print('Result:', k)

Output:

Enter: 1 2 3 4 5  
Result: 15

#7

## Enumerate()

- It allows you to create a for loop where we can iterate over a sequence and have both the items and their index in the loop.

syntax:

enumerate(sequence, start\_point)

Sequence → list or tuple or set of elements

start\_point → The starting index of the counter. default is 0

Example 1 (In list)

```
fruits = ['apple', 'Banana', 'orange']
```

```
for index, fruit in enumerate(fruit):  
    print(index, fruit)
```

output:  
0 apple  
1 Banana  
2 orange

Example 2 (In list) can also do in list comprehension

```
fruits = ['A', 'B', 'C']
```

```
for index, fruit in enumerate(fruit, start=1):  
    print(index, fruit)
```

output:  
1 A  
2 B  
3 C

Example 3 (In Dict)

```
fruits = {'a': 'apple', 'b': 'banana'}
```

```
for index, (key, value) in enumerate(fruits.items()):
```

```
    print(index, key, value)
```

output:  
0 a apple  
1 b banana  
23

#8

Sorted( )

- Used to return a new sorted list from the elements of any iterable. The original iterable is not modified, ie) original list is not modified. The function can sort the elements in ascending or descending order. ascending is default i.e) reverse=False

Syntax:

sorted (iterable, key=None, reverse=False/True)

iterable → list of elements to sort

key → a function that serves as a key for the sort comparison. default is none.

reverse → a boolean value, sorts in descending order if reverse = True.

Eg 1: (ascending)

num = [3, 4, 6, 7, 1]  
sor = sorted(num)  
Print(sor)

output :

[1, 3, 4, 6, 7]

Eg 2: (descending)

num = [3, 4, 6, 7, 1]  
sor = sorted(num, reverse=True)  
Print(sor)

output :

[7, 6, 4, 3, 1]

Eg 3: (key) len, max, min str.lower, abs, etc....

words = ['boy', 'girl', 'Horse']  
sor = sorted(words, key=len)  
Print(sor)

output :

['boy', 'girl', 'Horse']

# freecampcode

## Algorithms in Python

- helps solve problems efficiently
  - algorithms are specific set of instructions that we write to our computer to tell what to do
- what is algorithm

The current term of choice for a problem solving procedure... It is commonly used nowadays for the set of rules a machine (...especially a computer) follows to achieve a particular goal

It does not always apply to computer-mediated activity, however. The term may as accurately be used... [for] the steps followed in making a pizza or solving a Rubik's cube as for computer-powered data analysis

Algorithm[s]... [are] often paired with words specifying the activity for which a set of rules have been designed

A search algorithm, for example, is a procedure that determines what kind of information is retrieved from a large mass of data

An encryption An encryption algorithm is a set of rules by which information or messages are encoded so that unauthorized person cannot read them

1. Simple Recursive Algorithm
2. Algorithms within Data Structure
3. Divide & conquer
4. Greedy Algorithm
5. Dynamic programming

## 1. Simple Recursive Algorithm

Recursion  $\rightarrow$  function calls itself until the program terminates

LIFO - Last In First Out

(Eg: last number which goes in will be the first number to come out)  
Easy:

factorial

def fact(n):

if  $n == 1$ :

    return  $n$  # when  $n$  is 1, return 1

else:

    temp = fact( $n-1$ ) # recursive call

    return  $temp * n$  # multiply the result of fact( $n-1$ ) by  $n$   
    print(fact(5)) # return computed value

Output:

120

- here,
- the function has a base case that stops the recursion when  $n$  is 1, the function returns 1
  - If  $n$  is not 1 the function calls itself with the argument  $n-1$ . This continues until  $n$  is reduced to 1



break down the execution step by step

1. fact(5) calls fact(4)
2. fact(4) calls fact(3)
3. fact(3) calls fact(2)
4. fact(2) calls fact(1)
5. fact(1) returns 1

when fact(1) returns 1, the stack unwinds in reverse order [follows LIFO principle],

1. fact(2) returns  $1 * 2 = 2$
2. fact(3) returns  $2 * 3 = 6$
3. fact(4) returns  $4 * 6 = 24$
4. fact(5) returns  $24 * 5 = 120$

In LIFO,

- The last function call made is the first one to be resolved
- when a recursive function calls itself, each call is placed on top of the call stack
- once the base case is reached, the calls start returning values, beginning with most recent calls

Advanced -

Permutation



what is permutation?

it refers to the different ways in which a set of objects can be arranged. If you have  $n$  distinct objects, the total number of permutations of these objects is given by the factorial of  $n$  ( $n!$ )

Example with "ABCDE"

"ABCDE" has 5 distinct characters

$$5! = 5 \times 4 \times 3 \times 2 \times 1 = 120$$

so, there are 120 different ways to arrange the characters in "ABCDE"

Position-specific permutations

if you want to determine how many times a specific character appears in a specific position across all permutations.

For 'A' at the 0<sup>th</sup> index,

Fix 'A' at the 0<sup>th</sup> index and permute the remaining four characters (B,C,D,E)

$$4! = 4 \times 3 \times 2 \times 1 = 24$$

Examples  $\rightarrow$

# Algorithms within Data Structures

- Data structures are containers or collection of data they provide a means of managing small or large amounts of values for uses like Databases, Web traffic monitoring, user registrations, indexing (e.g. arrays)

1. Linear search
2. Binary search
3. Bubble sort
4. Insertion sort

## 1. Linear search

How it works,

1. start at beginning : Begins with the first element of the list
2. check each element : compare each element in the list with the target value
3. stop if found : if the target value is found, return the index of the element
4. continue if not found : if the target value is not found, move to the next element and repeat the process
5. End of list : if the end of the list is reached without finding the target, return a value indicating the target is not present

Example,

```
def linear_search(arr, target)
    for index in range(len(arr)):
        if arr[index] == target:
            return index
    return -1
```

linear-search([1, 2, 3, 4, 5], 2)

## 2. Binary Search

- It's an efficient algorithm for finding the position of a target value within a sorted list

How it works,

1. Start with the entire list: consider the whole list as the search interval
2. Find the middle element: calculate middle index of the current interval
3. compare the middle element with the target:
  - if the middle element is equal to the target, the search is complete, and index is returned
  - if the middle element is less than the target, search the right half of the interval
  - if the middle element is greater than the target, search the left half of the interval
4. Repeat: continue the process on the remaining half until the target is found or the interval is empty

2 Examples →

# Dealing with Complex Numbers

# Hacker Rank

@1

# Dealing with Complex Numbers

```
import math
```

```
class complex(object)
```

```
    def __init__(self, real, imaginary):
```

# constructor to initialize a complex number with real and imaginary parts

```
        self.real = real
```

```
        self.imaginary = imaginary
```

```
    def __add__(self, no):
```

# overloading the + operator to add two complex numbers

returns self.\_\_class\_\_(self.real + no.real, self.imaginary + no.imaginary)

```
    def __sub__(self, no):
```

# overloading the - operator to subtract two complex numbers

returns self.\_\_class\_\_(self.real - no.real, self.imaginary + no.imaginary)

```
    def __mul__(self, no):
```

# overloading the \* operator to multiply two complex numbers

r = (self.real \* no.real) - (self.imaginary \* no.imaginary)

i = (self.real \* no.imaginary) + (self.imaginary \* no.real)

# Formatting the result to the desired string

format with two decimal places

return "{:.2f}{}{:.2f}{}".format(r, "+" if i >= 0 else "-",

abs(i), "i")

```
    def __truediv__(self, no):
```

# overloading the / operator to divide two complex numbers

:

:

$$\text{div} = \text{no.real}^{**2} + \text{no.imaginary}^{**2}$$

$$\text{real-part} = ((\text{self}.real * \text{no.real}) + (\text{self}.imaginary * \text{no.imaginary})) / \text{div}$$

$$\text{imag-part} = ((\text{self}.imaginary * \text{no.real}) - (\text{self}.real * \text{no.imaginary})) / \text{div}$$

return self.\_\_class\_\_(real-part, imag-part)

def mod(self):

# Method to calculate the ~~no.~~ modulus of a complex number

$$mo = \text{pow}(\text{pow}(\text{self}.real, 2) + \text{pow}(\text{self}.imaginary, 2), 0.5)$$

return self.\_\_class\_\_(mo, 0)

def \_\_str\_\_(self):

# overloading the str() function to get the string representation of a complex numbers

if self.imaginary == 0 :

result = "% .2f + 0.00i" % (self.imaginary)

elif self.real == 0 :

if self.imaginary >= 0 :

result = "0.00 + % .2f i" % (self.imaginary)

else:

result = "0.00 - % .2f i" % (abs(self.imaginary))

elif self.imaginary > 0 :

result = "% .2f + % .2f i" % (self.real, self.imaginary)

else:

result = "% .2f - % .2f i" % (self.real, abs(self.imaginary))

)

return result

:

```

if __name__ == '__main__':
    # Taking two sets of inputs for complex numbers
    c = map(float, input().split())
    d = map(float, input().split())
    # Creating complex number instances with the input
    # values
    x = complex(*c)
    y = complex(*d)
    # performing operations and printing results
    # Each operation (+, -, *, /, modulus) is performed
    # and the result is converted to string format using
    print(*map(str, [x+y, x-y, x*y, x/y, x.mod(), y.mod()]))

```

Output:

2 1

5 6

} input

7.00 + 7.00i

-3.00 - 5.00i

4.00 + 17.00i

0.26 - 0.11i

2.24 + 0.00i

7.81 + 0.00i

## Question ?

gives two complex numbers, and you have to print the result of their addition, subtraction, multiplication, division and modulus operations.

The real and imaginary precision part should be correct up to two decimal places.

### Input format :

one line of input : the real and imaginary part of a number, separated by a space

### Output format :

for two complex numbers C and D

- $C + D$
  - $C - D$
  - $C * D$
  - $C / D$
  - $\text{mod}(C)$
  - $\text{mod}(D)$
- } output format

for complex numbers with non-zero real (A) and complex part (B). output should be  $A + Bi$ ,

Replace the plus symbol (+) with a minus symbol (-)  
when  $B < 0$

For complex numbers with a zero complex part,  $A + 0.00i$

For complex numbers where the real part is zero and complex part is non-zero,  $0.00 + Bi$

## Explanation

- **class Definition**: we define a class 'complex' to represent complex numbers.
- **constructor (`__init__`)**: This special method initializes each instance of the class with 'real' and 'imaginary' parts
- **operator overloading**:

operator overloading allows us to define custom behavior for operators ('+', '-','\*','/') when they are used with instances of the class. This makes instances of the class behave more like built-in types

- **Addition (`__add__`)**

How it works: returns a new instance of 'complex' where real & img parts (sums)

Purpose: To define how to add two complex numbers

Syntax: `def __add__(self, other):`

- **Subtraction (`__sub__`)**

Purpose: To define how to subtract two complex numbers

Syntax: `def __sub__(self, other):`

How it works: returns a new instance of 'complex' with real & img parts (subtracted)

• **Multiplication (`__mul__`)**

Purpose: to multiply two complex numbers

Formula:  $(a+bi) * (c+di) = (ac-bd) + (ad+bc)i$

Syntax: `def __mul__(self, other):`

- **Division (`__truediv__`)**

Purpose: to divide one complex number by another

Formula:  $(a+bi) / (c+di) = [(ac+bd) + (bc-ad)i] / (c^2+d^2)$

Syntax: `def __truediv__(self, other):`

How it works: calculates the denominator (div), and results by real & img

## → Modulus Method

Purpose: to calculate modulus (magnitude) of a complex number

formula:  $|a+bi| = \sqrt{a^2 + b^2}$

for sqrt we imported math

Syntax: `def __modulus__(self):`

How it works: it calculates the modulus and returns a new 'complex' object with the modulus as real and 0 as imaginary part

## → string representation (`__str__`)

Purpose is to provide string representation of the complex number

Syntax: `def __str__(self)`

How it works: it formats the real and imaginary parts to two decimal places and returns a string

## → performing operations.

Performs ('+', '-', '\*', '/', 'modulus'), converts the results to strings and prints them each on a newline

```
Print(*map(str,[x+y,x-y,x*y,x/y,x.mod(),y.mod()]))  
, sep = '\n')
```

## → `self.__class__.__(args)`

How it works: it refers to the class of the current instance, allowing the creation of a new instance of the same class

Why it is used: Ensures that the returned object is of the same class, even if the method is used in subclass

Syntax:

→ The operator overloading is invoked when the specific operation is performed on the objects of the class, not when the objects are created

when operator overloading happens?

operator overloading methods (such as `\_\_add\_\_`, ...) are called when the corresponding operation (`+`, ...) is performed on the instance of the class. It happens dynamically at runtime, exactly at the moment the operation is encountered in the code.

### Detailed explanation

#### 1. object creation

$$\begin{cases} x = \text{complex}(c) \\ y = \text{complex}(d) \end{cases}$$

- At this point, the constructor `\_\_init\_\_` is called for each instance, initializing their 'real' and 'imaginary' parts, no operator overloading is happened yet

#### 2. operations

$$\begin{cases} \text{result\_add} = x + y & [\text{calls or triggers } x.\text{-add-}()] \\ \text{result\_sub} = x - y & [\text{calls or } x.\text{-sub-}()] \\ \text{result\_mul} = x * y & [\text{calls or } x.\text{-mul-}() \quad \cancel{x.\text{-add-}()}] \\ \text{result\_div} = x / y & [\text{calls or } x.\text{-div-}() \quad \cancel{x.\text{-sub-}()}] \\ \text{result\_mod\_x} = x.\text{mod}() & [\text{calls or } x.\text{-mod-}() \quad \cancel{x.\text{-mul-}()}] \\ \text{result\_mod\_y} = y.\text{mod}() & [\text{calls or } y.\text{-mod-}() \quad \cancel{y.\text{-sub-}()}] \end{cases}$$

#### 3. string representation

$$\begin{cases} \text{print}(*\text{map}(\text{str}, [\text{result\_add}, \text{result\_sub}, \dots]), \text{sep}':') \end{cases}$$

The `\_\_str\_\_` method is called when the `str()` function is used, which happens inside the `map(str, ...)` call

Find a string

# HackerRank

@2

Find a String

def count\_substring(string, sub\_string):

count = 0 # initialize a counter to track the number of occurrence

lst1 = list(string) # convert the input string to a list of characters

st = [] # Initialize an empty list to store substrings of the same length as 'sub\_string'

# iterate over the string to create substrings of the same length as 'sub\_string'

for i in range(len(lst1) - len(sub\_string) + 1):

st.append([''.join(lst1[i:i + len(sub\_string)])])  
# create and add substrings to the list 'st'

# iterate over the list of substrings to count occurrences of

'sub\_string'

for j in st:

if j == sub\_string: # if the substring matches 'sub\_string'

count += 1 # increment the counter

else:

continue # continue to the next iteration if there's

return count # returns the total count of occurrences

if \_\_name\_\_ == '\_\_main\_\_':

string = input().strip() # read the input string from user

sub\_string = input().strip() # read the substring to search for from the user

count = count\_substring(string, sub\_string) # call the function with the inputs and store the result

Print(count) # print the result

- function 'count\_substring' takes two arguments 'string' and 'sub\_string'
- 'count' to keep track of the number of sub\_string in string
- The loop 'for i in range(len(lst1)-len(sub\_string)+1)' iterates over the indices of 'lst1' such that the slicing operation does not go out of bounds
- 'st.append([::join(lst1[i:i+len(sub\_string)])])' creates a substring starting at index 'i' with a length equal to 'sub\_string' and adds it to the list 'st'
- The loop 'for j in st' iterates over the list of generated substrings
- if the substring 'j' matches 'sub\_string', the count is incremented

# Time & space complexity

## Time & Space

# Big O notation

# Omega

# Theta

### Big O notation

- $O(1)$  — Constant
- $O(n)$  — Linear
- $O(\log N)$  — Logarithmic
- $O(\log(N \log N))$  — Linear Logarithmic
- $O(n^2)$  — Quadratic

#### $O(1)$

Constant time complexity means that the execution time of an algorithm or function does not depend on the size of the input. regardless of how big or small the input is, the time it takes to execute the function remains same.

Direc access: Typically involves operations that access a specific element or perform a simple computation

Predictable Performance: Since the execution time is constant, you can predict how long the operation will take, regardless of input size.

Scalability: Operations with  $O(1)$  complexity scale very well as the input size increases because their performance does not degrade.

### - $O(n)$

Linear time complexity means that the execution time of an algorithm or function increases linearly with the size of the input - if the input size doubles, the execution time also doubles.

Proportional Growth: The runtime grows proportionally with the size of the input.

Iteration over input: Typically involves iterating over elements in a data structure.

Predictable growth: you can predict how the runtime will increase as the input size increases.

# Best case

Eg:

$$L = [1, 2, 3]$$

$$a = 1$$

for  $i$  in  $L$ :

if  $i == a$ :

print ("True")

# Average case

$$L = [2, 1, 3]$$

$$a = 2$$

for  $i$  in  $L$ :

if  $i == a$ :

print ("True")

# Worst case

$$L = [3, 2, 1]$$

$$a = 1$$

for  $i$  in  $L$ :

if  $i == a$ :

print ("True")

## - $O(\log n)$

Logarithmic time complexity, means that the execution time of an algorithm increases logarithmically as the input size increases. In simpler terms, if you double the input size, the number of operations required only increases by a constant amount.

Slow growth: The runtime grows much more slowly compared to the input size.

Divide & conquer: Typically involves algorithms that divide the problem in half at each step.

- highly Scalable and handle large inputs efficiently

## - $O(n \log n)$

Linearithmic time complexity, means that the execution time of an algorithm grows in proportion to the input size  $n$  multiplied by the logarithm of the input size. This complexity often arises in algorithms that perform a combination of linear and logarithmic operations, such as divide-and-conquer strategies to each element.

Combination of Linear and Logarithmic: The runtime increases more slowly than quadratic time but faster than linear time.

common in Efficient sorting Algorithms: Many efficient sorting algorithms, like Merge Sort and Quick sort, have  $O(n \log n)$  complexity.

-  $O(n^2)$

quadratic time complexity, means that the execution time of an algorithm grows proportionally to the square of the input size. if the input size doubles, the execution time quadruples. This complexity often arises in algorithms that involve nested iterations over the

Rapid growth: the runtime grows much faster than linear times as the input size increase.

Nested Loops: typically involves algorithms with nested loops, where each element is compared with every other element.

## # Space Complexity

-  $O(1)$   $\rightarrow$  constant space

Definition - The amount of memory used is constant and does not depend on the input size

key points - Fixed memory usage, no additional space required based on input

-  $O(n)$   $\rightarrow$  linear space

Definition - The amount of memory used grows linearly with the input size

key points - memory usage is directly proportional to input size

-  $O(\log n)$   $\rightarrow$  Logarithmic space

Definition: The amount of memory used grows logarithmically as the input size increases.

Key points: Memory usage grows slowly, often related to recursive algorithms that divide the problem.

-  $O(n \log n)$   $\rightarrow$  Linearithmic space

Definition: the amount of memory used grows in proportion to the input size multiplied by the logarithm of the input size.

Key points: combination of linear and logarithmic growth, often seen in divide and conquer algorithms.

-  $O(n^2)$   $\rightarrow$  Quadratic Space complexity

Definition: The amount of memory used grows proportionally to the square of the input size.

Key points: Rapid growth in memory usage, often due to storing information in a two-dimensional structure.

escaryne vynut u 5

$O(1)$  → best

$O(\log n)$  ~~0 6 3 2 4~~

$O(n)$  5

$O(n \log n)$  15

$O(n^2)$  25

$O(n^3)$  125

$O(2^n)$  worst

# Array, Searching

## Array

- collection of data
- Same Data type
- Static in other language, Dynamic in Python

choosing data structure is based on these four operations

- ① Access
- ② insert
- ③ Delete
- ④ Search

- Elements of array is accessed by its index value

Time complexity

Syntax :

$O(1)$

array-name [index]

- inserting element in array



→ index  
→ elements

need to insert '3'

then

0	3	1	2

'1' and '2' moves to next index  
inserting '3' is first operation  
switching of '1' is second operation  
switching of '2' is third operation

length  
so, here  
 $O(h-i)$

$\geq O(4-1)$

$\geq O(3)$

Scene 2

1	1	1	2	6	.
---	---	---	---	---	---

inserting '5' here

so here  
 $\rightarrow O(n-i)$   
 $\rightarrow O(5-4)$   
 $\rightarrow O(1)$

done in constant time

- Deleting an element in array
- Scene 1

1	2	3	4	5
---	---	---	---	---

Deleting '1'

2	3	4	5
---	---	---	---

now '2', '3', '4', '5' will precede shifting to the

2	3	4	5
---	---	---	---

Scene 2

1	2	3	4	5
---	---	---	---	---

Deleting '5'

1	2	3	4
---	---	---	---

now →

$\left. \begin{array}{l} O(n-i) \\ O(5-4) \\ O(1) \end{array} \right\}$

- Searching an element in an array

6	7	2	5	1
---	---	---	---	---

X X X ✓

Searching '5'

→ best case

\* if '5' in index 0 or (first index) then, O(1)

\* if '5' is index 1 or 2 or 3

then it is not  $O(1)$  or  $O(n)$

?

↓ Average case

\* if '5' is last index (4), then it is  
 $O(n)$

↓ worst case

{ Now finally we should declare that  
 $O(n)$  is the time complexity of searching  
an element in array }

choosing array for a program

- program where there is a need to access element as it's time complexity is  $O(1)$  (if accessing is main need)
- program where inserting and deleting an element in last index as its time complexity is  $O(1)$  (if this is main need)  
should not choose array in best case or average case in inserting or deleting as the time complexity is  $O(n-i)$
- mostly in all data structure the searching time complexity is  $O(n)$ , so this is optional. In array

## Syntax of Array

import array as arr

variable=arr.array('type-code', [ ])

↓  
iterable

type-code

C type

Python type

Minimum  
Size is byte

'b'

signed char

int

1

'B'

unsigned char

int

1

'u'

Py\_unicode

unicode character

2

'h'

signed short

int

2

'H'

unsigned short

int

2

'i'

signed int

int

2

'I'

un " "

int

2

'l'

signed long

int

4

'L'

un " "

int

4

'q'

signed long long

int

8

'Q'

un " " "

int

8

'f'

float

float

4

'F'

double

float

8

# Searching Algorithms

## - Linear Search

equates each element from start to end with the search key, if it is equal returns index or success message otherwise (search out of the array) doesn't return anything

Time complexity is  $O(n)$ ,

The search key may be in best case or average case but we have to consider time complexity as worst case

[to store the returning index value the space complexity is  $O(1)$   
maybe we are not storing the index value, but minimum space complexity of a program is  $O(2)$ ]

## Example 1

array = [1, 2, 3, 4, 5]

a = 3

for i in range(len(array)):

if array[i] == a:

    Print(i)

else:

    break

    Print("not found")

## Example 2

array = [1, 2, 3, 4, 5]

a = 3

for i in array :

if i == a :

Print ("yes")

break

else :

Print ("not found")

## Binary Search

- Time complexity is  $O(\log n)$

Definition -

Binary search is a fast algorithm for finding a target value within a sorted array.

Steps :

1. Initialize pointers : set two pointers, low (starting at the beginning of the array) and high (starting at the end).

2. Calculate Midpoint : compute the midpoint  
 $mid = (low + high)/2$

3. Comparison :

if  $array[mid] == \text{target}$ , the target is found

if  $array[mid] < \text{target}$ , update low to  $mid + 1$

if  $array[mid] > \text{target}$ , update high to  $mid - 1$

Repeat until low exceeds high or target is found

Example:

nums = [1, 2, 3, 4, 5, 6, 7]  
key = 5  
l = 0

r = len(nums) - 1

while l <= r :

    mid = (l + r) // 2

    if ~~nums~~ nums[mid] == key :

        print(mid)

    elif ~~break~~ break

        nums[mid] < key :

            l = mid + 1

    elif ~~nums~~ nums[mid] > key :

        r = mid - 1

else :

    print("not found")

Reason for  $O(\log n)$  : - GPT

1. Initial intervals:

    • You start with an array of size (n)

2. Halving the Interval:

    • Each step of the binary search cuts the interval size in half

- In the first step, the interval is reduced from  $(n)$  to  $(n/2)$
- In the second step, it's reduced from  $(n/2)$  to  $(n/4)$
- This halving continues until the interval size becomes 1

### 3. Number of Halves:

- The number of times you can divide  $(n)$  by 2 until you get 1 is the logarithm base 2 of  $(n)$  denoted as  $(\log_2 n)$
- Mathematically, if you divide  $(n)$  by 2,  $(k)$  times, to get 1, then:  $\left[ \text{times} \backslash \text{left}(\backslash \text{proc}\{1\}\{2\} \backslash \text{right}) \right]$

$$^k = 2 \backslash \text{implies} \backslash \text{left}(\backslash \text{proc}\{1\}\{2\} \backslash \text{right})$$

$$^k = \backslash \text{frac}\{1\}\{2\} \backslash \text{implies} \\ k = \lfloor \log_2 n \rfloor$$

### 4. Iterations

- Thus, the number of iterations (or steps) required to reduce the interval to 1 (where you either find the target or conclude it isn't present) is  $(\log_2 n)$ .

## Recursion

## Recursion

- Function which calls itself
  - a logic written number of times can be replaced by recursion with the same logic
  - base condition is must or else infinite iteration will happen

for factorial (5)

- $\underline{5 \times 4 \times 3 \times 2 \times 1}$
- $5 \times \underline{4!}^n$
- $5 \times \underline{4 \times 3!}^n$
- $5 \times 4 \times \underline{3 \times 2!}^n$
- $\underline{5 \times 4 \times 3 \times 2 \times 1}$
- $5 \times 4 \times 3 \times 2 \times 1$



1 x 1!

To stop this  
we use base  
condition

as we know the  
value

value of 1! we break  
this and get..

as we know the value of  $1!$  we break this and return the value(1)

to find the logic, we have to find that, is there any relation between the sub problem

def fact(n):

if  $n == 1$ :

    returns 1

    returns

$n * \text{fact}(n-1)$

print(fact(5))



1.  $n$  will be 5 and the  $n$  changes to "fact( $n-1$ )" and again called with  $n=4$ .

2. now  $n$  will be 4 and then the  $n$  changes to  $(5 \times 4!)$  and again called with  $n=3$ , then the 4 will be stored in stack,  $(5 \times 4!)$

3. now  $n$  will be 3 and then the  $n$  changes to  $(4 \times 3!)$  and again called with  $n=2$ , then 3 will be stored in stack,  $(4 \times 3!)$

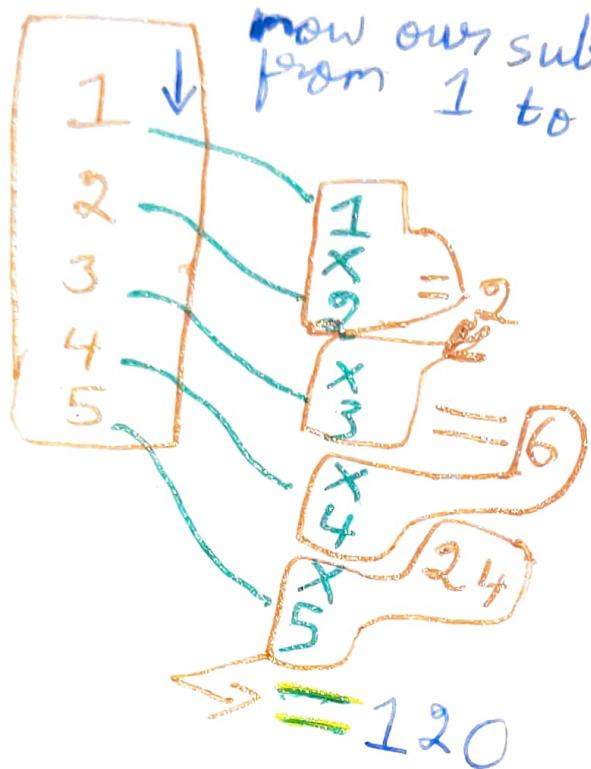
4. now  $n$  will be 2 and then the  $n$  changes to  $(3 \times 2!)$  and again called with  $n=1$  then 2 will be stored in stack,  $(3 \times 2!)$

5. Now, it goes to base condition "if  $n==1$ :" and returns the value of 1!

At step 5, instead of recursive function, base function works as ' $n == 1$ '

now, how stack works,

stored values



now our sub problem starts to multiply  
from 1 to 5