

## 1. Python Iterators

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**Iterators are objects that can be iterated upon. In this tutorial, you will learn how iterator works and how you can build your own iterator using `__iter__` and `__next__` methods.**

### What are iterators in Python?

Iterators are everywhere in Python. They are elegantly implemented within for loops, comprehensions, generators etc. but hidden in plain sight.

Iterator in Python is simply an [object](#) that can be iterated upon. An object which will return data, one element at a time.

Technically speaking, Python **iterator object** must implement two special methods, `__iter__()` and `__next__()`, collectively called the **iterator protocol**.

An object is called **iterable** if we can get an iterator from it. Most of built-in containers in Python like: [list](#), [tuple](#), [string](#) etc. are iterables.

The `iter()` function (which in turn calls the `__iter__()` method) returns an iterator from them.

### Iterating Through an Iterator in Python

We use the `next()` function to manually iterate through all the items of an iterator. When we reach the end and there is no more data to be returned, it will raise `StopIteration`. Following is an example.

```
# define a list
my_list = [4, 7, 0, 3]

# get an iterator using iter()
my_iter = iter(my_list)

## iterate through it using next()

#prints 4
print(next(my_iter))

#prints 7
print(next(my_iter))

## next(obj) is same as obj.__next__()

#prints 0
print(my_iter.__next__())

#prints 3
print(my_iter.__next__())

## This will raise error, no items left
next(my_iter)
```

---

A more elegant way of automatically iterating is by using the [for loop](#). Using this, we can iterate over any object that can return an iterator, for example list, string, file etc.

```
1. >>>for element inmy_list:
2. ...     print(element)
3. ...
4. 4
5. 7
6. 0
7. 3
```

### How for loop actually works?

As we see in the above example, the `for` loop was able to iterate automatically through the list.

In fact the `for` loop can iterate over any iterable. Let's take a closer look at how the `for` loop is actually implemented in Python.

```
1. for element initerable:
2.     # do something with element
```

Is actually implemented as.

```
1. # create an iterator object from that iterable
2. iter_obj = iter(iterable)
3.
4. # infinite loop
5. whileTrue:
6.     try:
7.         # get the next item
8.         element = next(iter_obj)
9.         # do something with element
10. exceptStopIteration:
11.     # if StopIteration is raised, break from loop
12. break
```

So internally, the `for` loop creates an iterator object, `iter_obj` by calling `iter()` on the iterable.

Ironically, this `for` loop is actually an infinite [while loop](#).

Inside the loop, it calls `next()` to get the next element and executes the body of the `for` loop with this value. After all the items exhaust, `StopIteration` is raised which is internally caught and the loop ends. Note that any other kind of exception will pass through.

### Building Your Own Iterator in Python

Building an iterator from scratch is easy in Python. We just have to implement the methods `__iter__()` and `__next__()`.

The `__iter__()` method returns the iterator object itself. If required, some initialization can be performed.

The `__next__()` method must return the next item in the sequence. On reaching the end, and in subsequent calls, it must raise `StopIteration`.

Here, we show an example that will give us next power of 2 in each iteration. Power exponent starts from zero up to a user set number.

```
class PowTwo:
    """Class to implement an iterator
    of powers of two"""

    def __init__(self, max = 0):
        self.max = max

    def __iter__(self):
        self.n = 0
        return self

    def __next__(self):
        if self.n <= self.max:
            result = 2 ** self.n
            self.n += 1
            return result
        else:
            raise StopIteration
```

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## 2 Python Recursion

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### What is recursion in Python?

Recursion is the process of defining something in terms of itself.

A physical world example would be to place two parallel mirrors facing each other. Any object in between them would be reflected recursively.

### Python Recursive Function

We know that in Python, a [function](#) can call other functions. It is even possible for the function to call itself. These type of construct are termed as recursive functions.

Following is an example of recursive function to find the factorial of an integer.

Factorial of a number is the product of all the integers from 1 to that number. For example, the factorial of 6 (denoted as 6!) is  $1*2*3*4*5*6 = 720$ .

```
# An example of a recursive function to  
# find the factorial of a number
```

```
def calc_factorial(x):  
    """This is a recursive function  
    to find the factorial of an integer"""  
  
    if x == 1:  
        return 1  
    else:  
        return (x * calc_factorial(x-1))
```

```
num = 4  
print("The factorial of", num, "is", calc_factorial(num))
```

In the above example, `calc_factorial()` is a recursive function as it calls itself.

When we call this function with a positive integer, it will recursively call itself by decreasing the number.

Each function call multiplies the number with the factorial of number 1 until the number is equal to one. This recursive call can be explained in the following steps.

```
1. calc_factorial(4)      # 1st call with 4  
2. 4 * calc_factorial(3)  # 2nd call with 3  
3. 4 * 3 * calc_factorial(2) # 3rd call with 2  
4. 4 * 3 * 2 * calc_factorial(1) # 4th call with 1  
5. 4 * 3 * 2 * 1 # return from 4th call as number=1  
6. 4 * 3 * 2 # return from 3rd call  
7. 4 * 6 # return from 2nd call  
8. 24 # return from 1st call
```

Our recursion ends when the number reduces to 1. This is called the base condition.

Every recursive function must have a base condition that stops the recursion or else the function calls itself infinitely.

### Advantages of Recursion

1. Recursive functions make the code look clean and elegant.
  2. A complex task can be broken down into simpler sub-problems using recursion.
  3. Sequence generation is easier with recursion than using some nested iteration.
- 

### Disadvantages of Recursion

1. Sometimes the logic behind recursion is hard to follow through.
2. Recursive calls are expensive (inefficient) as they take up a lot of memory and time.
3. Recursive functions are hard to debug.

## 3. Python Program to Display Fibonacci sequence Using Recursion

**In this program, you'll learn to display Fibonacci sequence using a recursive function.**

To understand this example, you should have the knowledge of following Python programming topics:

- Python for Loop
- Python Functions
- Python Recursion

A Fibonacci sequence is the integer sequence of 0, 1, 1, 2, 3, 5, 8....

The first two terms are 0 and 1. All other terms are obtained by adding the preceding two terms. This means to say the  $n$ th term is the sum of  $(n-1)^{\text{th}}$  and  $(n-2)^{\text{th}}$  term.

### Source Code

```
1. # Python program to display the Fibonacci sequence
2.
3. def recur_fibo(n):
4.     if n <= 1:
5.         return n
6.     else:
7.         return(recur_fibo(n-1) + recur_fibo(n-2))
8.
9. nterms = 10
10.
11. # check if the number of terms is valid
12. if nterms <= 0:
13.     print("Please enter a positive integer")
14. else:
15.     print("Fibonacci sequence:")
16.     for i in range(nterms):
17.         print(recur_fibo(i))
```

## Output

Fibonacci sequence:

0  
1  
1  
2  
3  
5  
8  
13  
21  
34

**Note:** To test the program, change the value of `nterms`.

In this program, we store the number of terms to be displayed in `nterms`.

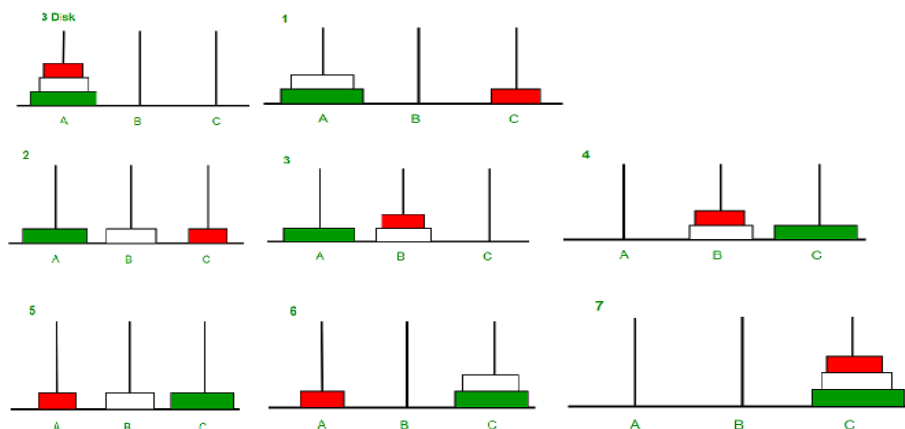
A recursive function `recur_fibo()` is used to calculate the *nth* term of the sequence. We use a `for` loop to iterate and calculate each term recursively.

Visit [here](#) to know more about [recursion in Python](#).

## 4. Python Program for Tower of Hanoi

Tower of Hanoi is a mathematical puzzle where we have three rods and *n* disks. The objective of the puzzle is to move the entire stack to another rod, obeying the following simple rules:

- 1) Only one disk can be moved at a time.
- 2) Each move consists of taking the upper disk from one of the stacks and placing it on top of another stack i.e. a disk can only be moved if it is the uppermost disk on a stack.
- 3) No disk may be placed on top of a smaller disk.



# Recursive Python function to solve tower of hanoi

```
def TowerOfHanoi(n , from_rod, to_rod, aux_rod):  
    if n == 1:  
        print("Move disk 1 from rod",from_rod,"to rod",to_rod)  
        return  
    TowerOfHanoi(n-1, from_rod, aux_rod, to_rod)  
    print("Move disk",n,"from rod",from_rod,"to rod",to_rod)  
    TowerOfHanoi(n-1, aux_rod, to_rod, from_rod)
```

# Driver code

n =4

TowerOfHanoi(n, 'A', 'C', 'B')

# A, C, B are the name of rods

Output:

```
Move disk 1 from rod A to rod B  
Move disk 2 from rod A to rod C  
Move disk 1 from rod B to rod C  
Move disk 3 from rod A to rod B  
Move disk 1 from rod C to rod A  
Move disk 2 from rod C to rod B  
Move disk 1 from rod A to rod B  
Move disk 4 from rod A to rod C  
Move disk 1 from rod B to rod C  
Move disk 2 from rod B to rod A  
Move disk 1 from rod C to rod A  
Move disk 3 from rod B to rod C  
Move disk 1 from rod A to rod B  
Move disk 2 from rod A to rod C  
Move disk 1 from rod B to rod C
```

## Sorting, searching and algorithm analysis

### Introduction

We have learned that in order to write a computer program which performs some task we must construct a suitable algorithm. However, whatever algorithm we construct is unlikely to be unique – there are likely to be many possible algorithms which can perform the same task. Are some of these algorithms in some sense better than others? Algorithm analysis is the study of this question.

In this chapter we will analyse four algorithms; two for each of the following common tasks:

- *sorting*: ordering a list of values
- *searching*: finding the position of a value within a list

Algorithm analysis should begin with a clear statement of the task to be performed. This allows us both to check that the algorithm is correct and to ensure that the algorithms we are comparing perform the same task.

Although there are many ways that algorithms can be compared, we will focus on two that are of primary importance to many data processing algorithms:

- *time complexity*: how the number of steps required depends on the size of the input
- *space complexity*: how the amount of extra memory or storage required depends on the size of the input

### Sorting algorithms

The sorting of a list of values is a common computational task which has been studied extensively. The classic description of the task is as follows:

Given a *list of values* and a function that *compares two values*, order the values in the list from smallest to largest.

The values might be integers, or strings or even other kinds of objects. We will examine two algorithms:

- *Selection sort*, which relies on repeated *selection* of the next smallest item
- *Merge sort*, which relies on repeated *merging* of sections of the list that are already sorted



Other well-known algorithms for sorting lists are *insertion sort*, *bubble sort*, *heap sort*, *quicksort* and *shell sort*.

There are also various algorithms which perform the sorting task for restricted kinds of values, for example:

- *Counting sort*, which relies on the values belonging to a small set of items
- *Bucket sort*, which relies on the ability to map each value to one of a small set of items
- *Radix sort*, which relies on the values being sequences of digits

If we restrict the task, we can enlarge the set of algorithms that can perform it. Among these new algorithms may be ones that have desirable properties. For example, *Radix sort* uses fewer steps than any generic sorting algorithm.

Algorithm	Best case	Expected	Worst case
Selection sort	$O(N^2)$	$O(N^2)$	$O(N^2)$
Merge sort	$O(N \log N)$	$O(N \log N)$	$O(N \log N)$
Linear search	$O(1)$	$O(N)$	$O(N)$
Binary search	$O(1)$	$O(\log N)$	$O(\log N)$

What does  $O(1)$  mean? It means that the cost of an algorithm is *constant*, no matter what the value of  $N$  is. For both these search algorithms, the best case scenario happens when the first element to be tested is the correct element – then we only have to perform a single operation to find it.

In the previous table, big  $O$  notation has been used to describe the *time complexity* of algorithms. It can also be used to describe their *space complexity* – in which case the cost function represents the number of units of space required for storage rather than the required number of operations. Here are the space complexities of the algorithms above (for the worst case, and excluding the space required to store the input):

Algorithm	Space complexity
Selection sort	O(1)
Merge sort	O(N)
Linear search	O(1)
Binary search	O(1)

## 5. Python | Merge list elements

Sometimes, we require to merge some of the elements as single element in the list. This is usually with the cases with character to string conversion. This type of task is usually required in the development domain to merge the names into one element. Let's discuss certain ways in which this can be performed.

### Method #1 :

#### Using join()+ListSlicing

The join function can be coupled with list slicing which can perform the task of joining each character in a range picked by the list slicing functionality.

```
# Python3 code to demonstrate
```

```
# merging list elements
```

```
# using join() + list slicing
```

```
# initializing list
```

```
test_list = ['T', 'L', 'O', 'V', 'E', 'G', 'F', 'G']
```

```
# printing original list
```

```
print("The original list is : "+str(test_list))
```

```
# using join() + list slicing
```

```
# merging list elements
```

```
test_list[5: 8] =['.join(test_list[5: 8])]
```

```
# printing result
```

```
print("The list after merging elements : "+ str(test_list))
```

**Output:**

```
The original list is : ['T', 'L', 'O', 'V', 'E', 'G', 'F', 'G']
```

```
The list after merging elements : ['T', 'L', 'O', 'V', 'E', 'GFG']
```

**Method #2 : Using reduce() + lambda + list slicing**

The task of joining each element in a range is performed by reduce function and lambda. reduce function performs the task for each element in the range which is defined by the lambda function. It works with Python2 only

```
# Python code to demonstrate
```

```
# merging list elements
```

```
# using reduce() + lambda + list slicing
```

```
# initializing list
```

```
test_list =['T', 'L', 'O', 'V', 'E', 'G', 'F', 'G']
```

```
# printing original list
```

```
print("The original list is : "+str(test_list))
```

```
# using reduce() + lambda + list slicing
```

```
# merging list elements
```

```
test_list[5: 8] =[reduce(lambdai, j: i +j, test_list[5: 8])]
```

```
# printing result
```

```
print("The list after merging elements : "+ str(test_list))
```

**Output:**

```
The original list is : ['T', 'L', 'O', 'V', 'E', 'G', 'F', 'G']
```

```
The list after merging elements : ['T', 'L', 'O', 'V', 'E', 'GFG']
```

## 6. Linear Search

**Problem:** Given an array `arr[]` of `n` elements, write a function to search a given element `x` in `arr[]`.

### Examples :

Input :`arr[] = {10, 20, 80, 30, 60, 50, 110, 100, 130, 170}`

`x = 110;`

Output : 6

Element `x` is present at index 6

Input :`arr[] = {10, 20, 80, 30, 60, 50, 110, 100, 130, 170}`

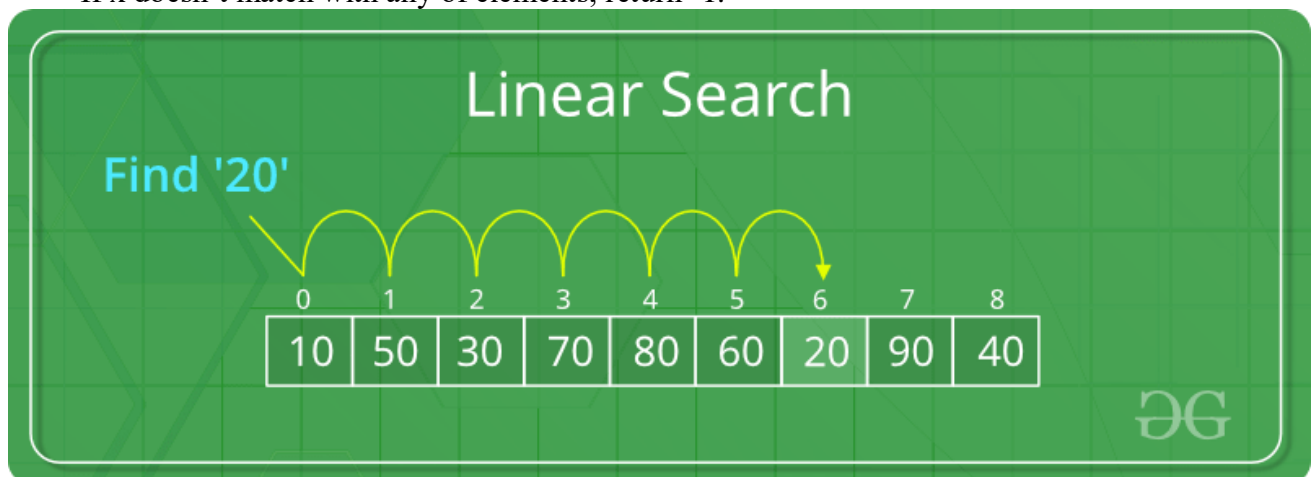
`x = 175;`

Output : -1

Element `x` is not present in `arr[]`.

A simple approach is to do **linear search**, i.e

- Start from the leftmost element of `arr[]` and one by one compare `x` with each element of `arr[]`
- If `x` matches with an element, return the index.
- If `x` doesn't match with any of elements, return -1.



### Example:

```
# Python3 code to linearly search x in arr[].  
# If x is present then return its location,
```

```

# otherwise return -1

def search(arr, n, x):

    for i in range(0, n):
        if(arr[i] == x):
            return i;
    return -1;

# Driver Code
arr = [ 2, 3, 4, 10, 40];
x = 10;
n = len(arr);
result = search(arr, n, x)
if(result == -1):
    print("Element is not present in array")
else:
    print("Element is present at index", result);

```

**Output:**

Element is present at index 3

The time complexity of above algorithm is  $O(n)$ .

Linear search is rarely used practically because other search algorithms such as the binary search algorithm and hash tables allow significantly faster searching comparison to Linear search.

## 7. Binary Search

Given a sorted array `arr[]` of `n` elements, write a function to search a given element `x` in `arr[]`.

A simple approach is to do **linear search**. The time complexity of above algorithm is  $O(n)$ . Another approach to perform the same task is using Binary Search.

**Binary Search:** Search a sorted array by repeatedly dividing the search interval in half. Begin with an interval covering the whole array. If the value of the search key is less than the item in the middle of the interval, narrow the interval to the lower half. Otherwise narrow it to the upper half. Repeatedly check until the value is found or the interval is empty.

## Example

# Binary Search

	0	1	2	3	4	5	6	7	8	9
Search 23	2	5	8	12	16	23	38	56	72	91
	L=0	1	2	3	M=4	5	6	7	8	H=9
23 > 16 take 2 <sup>nd</sup> half	2	5	8	12	16	23	38	56	72	91
	0	1	2	3	4	L=5	6	M=7	8	H=9
23 > 56 take 1 <sup>st</sup> half	2	5	8	12	16	23	38	56	72	91
	0	1	2	3	4	L=5, M=5	H=6	7	8	9
Found 23, Return 5	2	5	8	12	16	23	38	56	72	91

The idea of binary search is to use the information that the array is sorted and reduce the time complexity to  $O(\log n)$ .

We basically ignore half of the elements just after one comparison.

1. Compare  $x$  with the middle element.
2. If  $x$  matches with middle element, we return the mid index.
3. Else If  $x$  is greater than the mid element, then  $x$  can only lie in right half subarray after the mid element. So we recur for right half.
4. Else ( $x$  is smaller) recur for the left half.

### Recursive implementation of Binary Search

# Python Program for recursive binary search.

# Returns index of  $x$  in arr if present, else -1  
def binarySearch (arr, l, r, x):

# Check base case  
if r >= l:

mid = l + (r - l) / 2

# If element is present at the middle itself  
if arr[mid] == x:  
return mid

# If element is smaller than mid, then it  
# can only be present in left subarray  
elif arr[mid] > x:  
return binarySearch(arr, l, mid - 1, x)

```

        # Else the element can only be present
        # in right subarray
        else:
            return binarySearch(arr, mid + 1, r, x)

    else:
        # Element is not present in the array
        return -1

# Test array
arr = [ 2, 3, 4, 10, 40]
x = 10

# Function call
result = binarySearch(arr, 0, len(arr)-1, x)

if result != -1:
    print "Element is present at index % d"%result
else:
    print "Element is not present in array"

```

### Output :

```
Element is present at index 3
```

## Iterative implementation of Binary Search

```

# Python code to implement iterative Binary
# Search.

```

```

# It returns location of x in given array arr
# if present, else returns -1
def binarySearch(arr, l, r, x):

```

```

    while l <= r:

```

```

        mid = l + (r - l) // 2;

```

```

        # Check if x is present at mid
        if arr[mid] == x:
            return mid

```

```

        # If x is greater, ignore left half
        elif arr[mid] < x:
            l = mid + 1

```

```

        # If x is smaller, ignore right half
        else:
            r = mid - 1

```

```

# If we reach here, then the element
# was not present
return -1

# Test array
arr = [2, 3, 4, 10, 40]
x = 10

# Function call
result = binarySearch(arr, 0, len(arr)-1, x)

if result != -1:
    print("Element is present at index % d"%result)
else:
    print("Element is not present in array")

```

### Output :

Element is present at index 3

### Time

The time complexity of Binary Search can be written as

$$T(n) = T(n/2) + c$$

### Complexity:

The above recurrence can be solved either using Recurrence Tree method or Master method.

It falls in case II of Master Method and solution of the recurrence is .

**Auxiliary Space:**  $O(1)$  in case of iterative implementation. In case of recursive implementation,  $O(\log n)$  recursion call stack space.

## 8. Selection Sort Algorithm

**In this tutorial, you will learn how selection sort works. Also, you will find working examples of selection sort in C, C++, Java and Python.**

Selection sort is an algorithm that selects the smallest element from an unsorted list in each iteration and places that element at the beginning of the unsorted list.

### How Selection Sort Works?

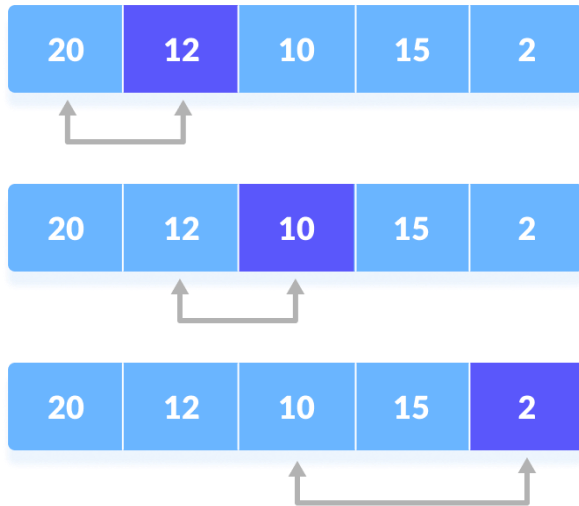
1. Set the first element as minimum.



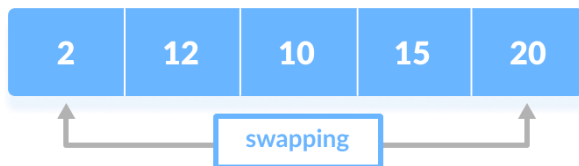


2. Compare **minimum** with the second element. If the second element is smaller than **minimum**, assign **second element** as **minimum**.

Compare **minimum** with the third element. Again, if the third element is smaller, then assign **minimum** to the third element otherwise do nothing. The process goes on until the last element.

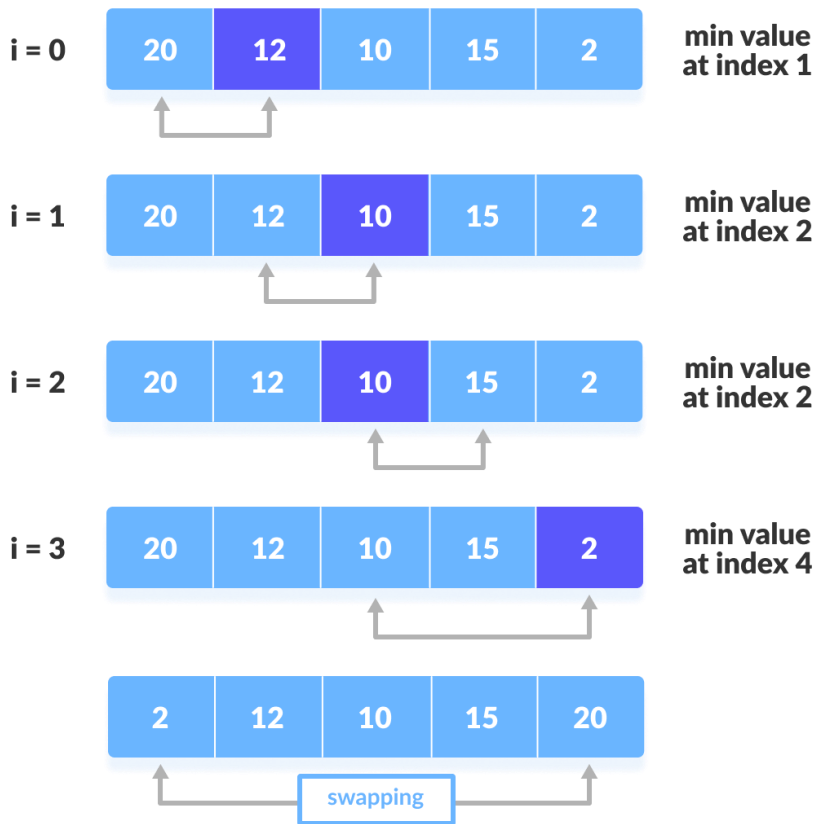


3. After each iteration, **minimum** is placed in the front of the unsorted list.

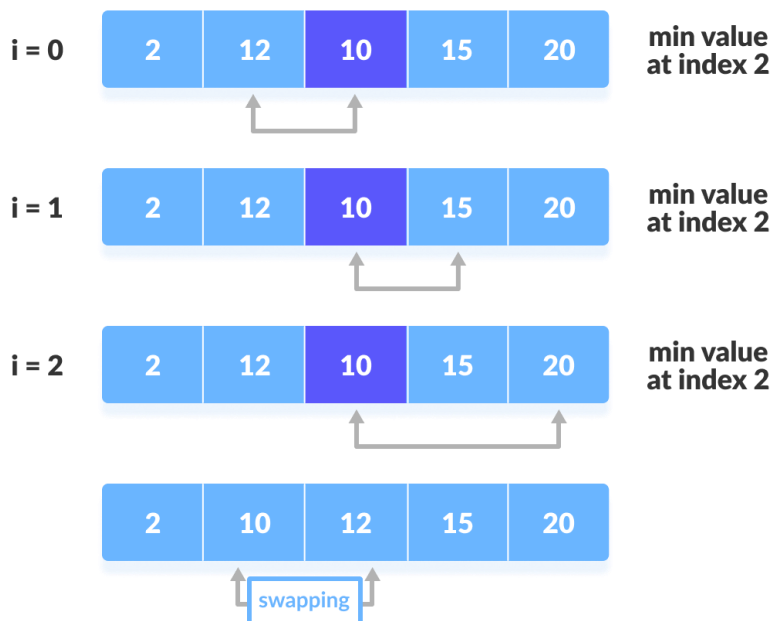


4. For each iteration, indexing starts from the first unsorted element. Step 1 to 3 are repeated until all the elements are placed at their correct positions.

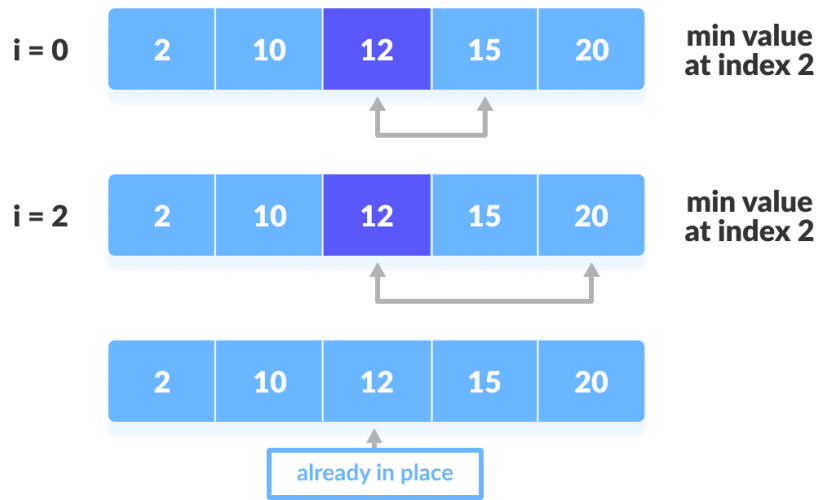
step = 0



step = 1



step = 2



step = 3



### Selection Sort Algorithm

1. selectionSort(array, size)
2.    repeat (size - 1) times
3.    set the first unsorted element as the minimum
4.    for each of the unsorted elements
5.    if element < currentMinimum
6.    set element as new minimum
7.    swap minimum with first unsorted position
8. endselectionSort

- Python

```
1. # Selection sort in Python
2.
3. def selectionSort(array, size):
4.     for step in range(size):
5.         min_idx = step
6.         for i in range(step + 1, size):
7.
8.             # To sort in descending order, change > to < in this line.
9.
10.            if array[i] < array[min_idx]:
11.                min_idx = i
12.
13.            (array[step], array[min_idx]) = (array[min_idx], array[step])
14.
15.
16. data = [-2, 45, 0, 11, -9]
17. size = len(data)
18. selectionSort(data, size)
19. print('Sorted Array in Ascending Order:\n')
20. print(data)
```

### Complexity

Cycle	Number of Comparison
1st	(n-1)
2nd	(n-2)
3rd	(n-3)
...	...
last	1

**Number of comparisons:**  $(n-1) + (n-2) + (n-3) + \dots + 1 = n(n-1)/2$  nearly equals to  $n^2$

**Complexity** =  $O(n^2)$

Also, we can analyze the complexity by simply observing the number of loops. There are 2 loops so the complexity is  $n*n = n^2$ .

### Time Complexities:

**WorstCaseComplexity:**  $O(n^2)$

If we want to sort in ascending order and the array is in descending order then, the worst case occurs.

- **BestCaseComplexity:**  $O(n^2)$

It occurs when the array is already sorted

- **AverageCaseComplexity:**  $O(n^2)$

It occurs when the elements of the array are in jumbled order (neither ascending nor descending).

The time complexity of selection sort is the same in all cases. At every step, you have to find the minimum element and put it in the right place. The minimum element is not known until the end of the array is not reached.

### Space Complexity:

Space complexity is  $O(1)$  because an extra variable `temp` is used.

### Selection Sort Applications

The selection sort is used when:

- small list is to be sorted
- cost of swapping does not matter
- checking of all the elements is compulsory
- cost of writing to a memory matters like in flash memory (number of writes/swaps is  $O(n)$  as compared to  $O(n^2)$  of bubble sort)

## 9. Python Program for Merge Sort

### Merge Sort

Like QuickSort, Merge Sort is a Divide and Conquer algorithm. It divides input array in two halves, calls itself for the two halves and then merges the two sorted halves. **The merge () function** is used for merging two halves. The `merge(arr, l, m, r)` is key process that assumes that `arr[l..m]` and `arr[m+1..r]` are sorted and merges the two sorted sub-arrays into one

**MergeSort(arr[], l, r)**

If  $r > l$

1. Find the middle point to divide the array into two halves:

```

        middle m = (l+r)/2
    2. Call mergeSort for first half:
        Call mergeSort(arr, l, m)
    3. Call mergeSort for second half:
        Call mergeSort(arr, m+1, r)
    4. Merge the two halves sorted in step 2 and 3:
        Call merge(arr, l, m, r)

```

Merge Sort is a **Divide and Conquer** algorithm. It divides input array in two halves, calls itself for the two halves and then merges the two sorted halves. **The merge () function** is used for merging two halves. The merge(arr, l, m, r) is key process that assumes that arr[l..m] and arr[m+1..r] are sorted and merges the two sorted sub-arrays into one.

### # Python program for implementation of MergeSort

```

# Merges two subarrays of arr[].
# First subarray is arr[l..m]
# Second subarray is arr[m+1..r]
def merge(arr, l, m, r):
    n1 = m - l + 1
    n2 = r - m

    # create temp arrays
    L = [0] * (n1)
    R = [0] * (n2)

    # Copy data to temp arrays L[] and R[]
    for i in range(0, n1):
        L[i] = arr[l + i]

    for j in range(0, n2):
        R[j] = arr[m + 1 + j]

    # Merge the temp arrays back into arr[l..r]
    i = 0 # Initial index of first subarray
    j = 0 # Initial index of second subarray

```

```
k = l    # Initial index of merged subarray
```

```
while i < n1 and j < n2 :
```

```
    if L[i] <= R[j]:
```

```
        arr[k] = L[i]
```

```
        i += 1
```

```
    else:
```

```
        arr[k] = R[j]
```

```
        j += 1
```

```
    k += 1
```

```
# Copy the remaining elements of L[], if there
```

```
# are any
```

```
while i < n1:
```

```
    arr[k] = L[i]
```

```
    i += 1
```

```
    k += 1
```

```
# Copy the remaining elements of R[], if there
```

```
# are any
```

```
while j < n2:
```

```
    arr[k] = R[j]
```

```
    j += 1
```

```
    k += 1
```

```
# l is for left index and r is right index of the
```

```
# sub-array of arr to be sorted
```

```
def mergeSort(arr, l, r):
```

```
    if l < r:
```

```
        # Same as (l+r)/2, but avoids overflow for
```

```
        # large l and h
```

```
        m = (l + (r - 1)) // 2
```

```
# Sort first and second halves
mergeSort(arr, l, m)
mergeSort(arr, m+1, r)
merge(arr, l, m, r)

# Driver code to test above
arr =[12, 11, 13, 5, 6, 7]
n =len(arr)
print("Given array is")
for i in range(n):
    print("%d"%arr[i]),

mergeSort(arr,0,n-1)
print("\n\nSorted array is")
for i in range(n):
    print("%d"%arr[i]),

# This code is contributed by MohitKumra
```

**Output:**

Given array is

12 11 13 5 6 7

Sorted array is

5 6 7 11 12 13