

① Master Theorem

Let $T(n)$ be an eventually non decreasing function that satisfies the recurrence relation.

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n), \quad a = b^k \quad (k = 1, 2, 3, \dots) \text{ and } T(1) = c$$

where $a \geq 1, b \geq 2, c > 0$

$$f(n) = \Theta(n^k \log^p n)$$

$$T(n) \in \begin{cases} \text{case 1:} \\ \text{if } \log_b a > k \text{ then } \Theta(n^{\log_b a}) \\ \text{case 2:} \\ \text{if } \log_b a = k \\ \quad \text{if } p > -1 \quad \Theta(n^k \log^{p+1} n) \\ \quad \text{if } p = -1 \quad \Theta(n^k \log \log n) \\ \quad \text{if } p < -1 \quad \Theta(n^k) \\ \text{case 3:} \\ \text{if } \log_b a < k \quad \text{if } p \geq 0 \quad \Theta(n^k \log^p n) \\ \quad \text{if } p < 0 \quad \Theta(n^k) \end{cases}$$

a) $a=2, b=4, k=p=\frac{1}{2}$ $\frac{a}{b^k} = \frac{2}{4^{1/2}} = 1$ and $p > -1$ So $T(n) = \Theta(n^{\log_4 2} \cdot \log^{(1/2)+1} n) = \Theta(n^{1/2} \log^{3/2} n)$

b) $a=9, b=3, k=2, p=0$ $a > b^k$ and $p > -1$ So $T(n) = \Theta(n^{\log_3 9} \cdot \log^{0+1} n) = \Theta(n^2 \log n)$

c) $a=\frac{1}{2}$ is not ≥ 1 so master theorem cannot be applied.

d) $a=5, b=2, k=0, p=1$
 $a > b^k$ so $T(n) = \Theta(n^{\log_2 5})$

e) Exponentials doesn't fit with master theorem so cannot be solved.

f) $a=7, b=4, k=1, p=1$
 $a > b^k$ so $T(n) = \Theta(n^{\log_4 7})$

g) $a=2, b=3, k=-1$
 k is not ≥ 0 so cannot be solved with master theorem.

h) $a=\frac{2}{5}, b=5, k=5, p=0$ $a \neq 1$ So cannot be solved with master theorem.

② for i in range $(1, \ln(A))$:
 element = $A[i]$
 # Start from 2nd element of array, continue until last.
 # Saving element in i th index to variable so when
 # array later changes, it will stay the same.
 $j = i - 1$
 while ($j > 0$ and $A[j] > \text{element}$):
 $A[j+1] = A[j]$
 $j = j - 1$
 # Every time starting j with $i-1$ so,
 # So we can compare with previous element
 # while it's bigger than element.
 # if comparison holds true, move element to left
 # Put element to it's right place.
 $A[j+1] = \text{element}$

- 3 2 6 1 4 5
- 2 3 6 1 4 5
- 2 5 1 6 4 5
- 2 1 3 6 4 5
- 1 2 3 6 4 5
- 1 2 3 4 6 5
- 1 2 3 4 5 6

3

- a)
- i. Linked list holds first element. So, no iteration needed. $O(1)$
Array elements can be accessed through index so $O(1)$.
 - ii. If linked list doesn't hold it's tail as pointer we need to iterate through all. So $O(n)$
Array elements are indexed. So doesn't matter where, it's $O(1)$.
 - iii. Even element is in the middle, accessing will be iterating $\frac{n}{2}$ times for linked list, so $O(\frac{n}{2}) = O(n)$
Array elements are indexed. So doesn't matter where, it's $O(1)$.
 - iv. For linked list, we need to change the pointer of head element to new head without losing other data pointers. $O(1)$
For array, we need to create a new array/move every element so $O(n)$.
 - v. For linked list, we need to iterate until the end, then add. It's $O(n)$.
For array, because no extra space allocated, we need to create a new array, move all and then add. So $O(n)$
 - vi. For linked list, we need to iterate half of list, so $O(\frac{n}{2}) \Rightarrow O(n)$
For array, we need to create a new array, move elements until middle, then add new element, and then add rest of it. So $O(n)$
 - vii. For linked list, moving head to second element and deallocating the first is enough. So, $O(1)$
For array, creating a new array and copying from one to other will take $O(n)$ time.
 - viii. For linked list, moving iterator till the end, then deallocating is enough so it's $O(n)$.
For array we can just ignore last element. But forgetting array without the last element, we need to move every element except last. $O(n)$
 - ix. Both similar to viii, we need to iterate over elements so it will take $O(n)$ for array and linked list.

b) for extra space needed,

- i. No need extra space.
- ii. Linked list, we need to create an iterator size of pointer for array nothing needed.
- iii. Linked list needs an iterator to create, for array nothing needed.
- iv. Linked list - nothing needed; array - we need to create an array size of n .
- v. Linked list - one element needed; array - n element needed.
- vi. Linked list - one element needed; array - n element needed.
- vii. Linked list - nothing is needed; array - n element needed.
- viii. Linked list - nothing is needed; array - n element is needed.
- ix. Linked list - nothing is needed; array - n element is needed.

④ def BinaryTreeToBST (root, n):

if (root is None): # checking if tree is empty
return

tempArray = []

inorderStorer (root, tempArray) # Storing values in orderly in an array

tempArray.sort() # Sorting the array (python quicksort)

convertToBST (tempArray, root) # Sending from array to BST in orderly

def InorderStorer (root, stored):

if root is None:

return

inorderStorer (root.left, stored)

stored.append (root.data)

inorderStorer (root.right, stored)

def convertToBST (tempArray, root):

if root is None:

return # Checking if value to add empty recursively

convertToBST (tempArray, root.left) # First performing this operation for left side

root.data = tempArray[0] # Add first value to BST, then remove value

tempArray.pop (0) # from array

convertToBST (tempArray, root.right) # Perform this for right side

Time Complexities

- In orderly Storing all n values will take $O(n)$ time. (every case)
- In the worst case, quicksort will sort using pivot which is smallest or largest element of the array. This will happen when input array is already sorted & pivot will be first or last element. So it will take $O(n^2)$ time. At the end, in orderly copying array elements to tree nodes will take $O(n)$ time. (every case)
 $O(n + n^2 + n) \Rightarrow O(n^2)$
 - In the best case, when we partitioned the array to sort, when they are very evenly balanced e.g. their size are equal. That makes a balanced binary tree for quicksort. Because of binary tree have height of $(\log n)$, time complexity will be $O(n \log n)$
 $O(n + n \log n + n) = O(n \log n)$
 - In the average case, because average case compairs required by is recurrence relation, $T(n) = O(n) + 2 * T(\frac{n}{2})$, If we use master theorem, because a is 2, b is 2 $O(n^{\log_2 2}) = O(n)$. Every time $\frac{n}{2}$ is makes $\log n$ so at the end $O(n \log n)$.
- $O(n + n \log n + n) = O(n \log n)$.
- \uparrow
smaller than $n \log n$ so we can ignore.


```
def sort(array, low, high) : # Quicksort
```

```
    if low < high:
```

```
        # Find pivot element and smaller will be put on left
        # greater will be put in right
```

```
        part = partition(array, low, high) # Have partition on array
```

```
        sort(array, low, part - 1)
```

```
        # Operation for left side of pivot
```

```
        sort(array, part + 1, high)
```

```
        # Operation for right side of pivot
```

```
def partition(array, low, high) :
```

```
    pivot = array[high]
```

```
    # Element at the right end will be pivot
```

```
    i = low - 1
```

```
    # bigger element pointer
```

```
    for j in range(low, high) :
```

```
        if (array[j] <= pivot):
```

```
            i = i + 1
```

```
            # Swap smaller with bigger if found
```

```
            # but first increment i number
```

```
        (array[i], array[j]) = (array[j], array[i])
```

```
    (array[i+1], array[high]) = (array[high], array[i+1])
```

```
    # Swapping pivot with bigger element
```

```
    return i+1
```

```
    # Return end of partition value.
```


⑤ def pairfinder():
 array = [5, 15, 23, 18, 7, 22]
 x = 5

set = set()

for i in array:

if abs(i-x) in set:
 print(i, " and ", i-x)

if abs(i+x) in set:
 print(i+x, " and ", i)

set.add(i)

Example array

Example x

Creating empty set so every element

will be stored there to compare

with i+x and i-x's absolutes.

For every (element-x), check

if that value is already

seen so we can print a

valid pair

⑥

a) True.

Because root node value also changes other values' insertion place. (If smaller than root, it will be at left; if bigger, it will be at right.) Every added item will depend on its root for left-hand/right-hand place.

b) True

If tree is skewed, then we will go to one side only, so instead $\log n$, we will get $O(n)$.

c) True

Array's indexes are obvious. No need to iterate to compare any element of array with other element of array.

d) ^{false}

If linkedlist is a singly linked list and we want to examine the last node, we need to perform $n-1$ operations to access next pointer, Memory allocation is not ambiguous so it will take $O(n)$ to find middle element.

e) false.

worst case occurs when array is sorted in reverse order. Because, there are 2 loops. Inner loop iterates from current position to the first element, outer will iterate no matter what so it will take $O(n^2)$.