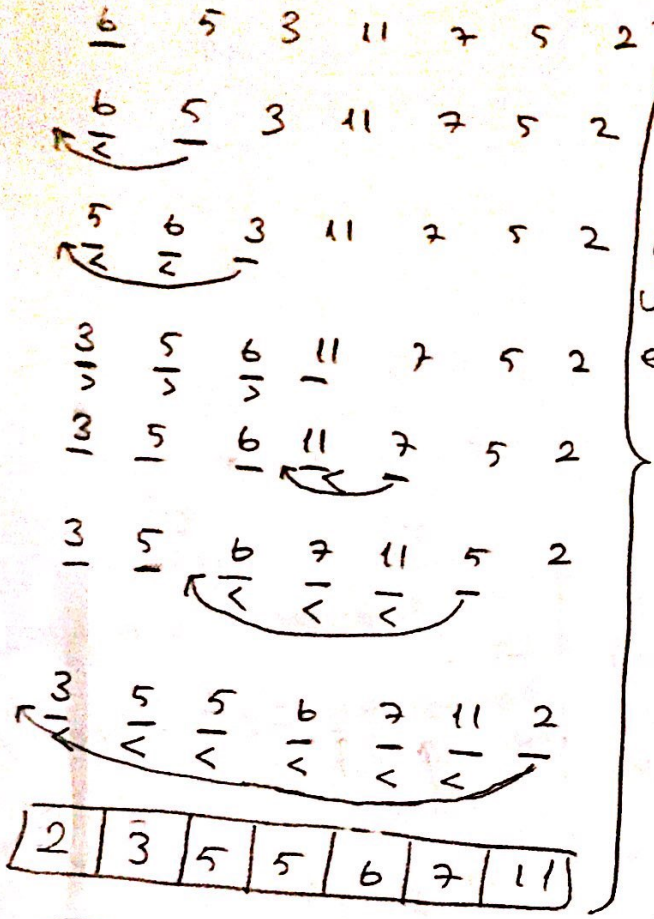
① 6, 5, 3, 11, 7, 5, 2



If the key elements is smaller than its predecessor, compare it to the elements before. Move the greater elements one position up to make space for the swapped element.

(2.a)

```
function(int n) {
    if(n==1)
        return;
    for(int i=1; i<=n; i++) {
        for(int j=1; j<=n; j++) {
            printf("*");
            break;
        }
    }
}
```

Best case $\Omega(1)$
=> worst case $O(n)$

Best case complexity

When the n is 1, function don't run loop just return. "return" is one process, so the function is $\Omega(1)$

Worst case Complexity

The second loop will run only once because there is "break" in second loop code scope. The first loop will run n times. Because they are nested, $O(1) \times O(n) = O(n)$

(2.5)

```
void function (int n) {
    int count = 0;
    for(int i = n/3; i <= n; ++i)
        for(int j = 1; j+n/3 <= n; j++)
            for(int k=1; k<=n; k = k*3)
                ++count;
}
```

First loop

The loop runs $(n - \frac{n}{3})$ times

Second Loop

The loop runs x times. $\frac{n}{3}$ is constant number because the loop takes this number a top loop.

$(n - \frac{n}{3}) = x$

Third Loop

The loop runs $\log_3 n$ times.

Proof

Suppose that third loop runs x times.

1, 3, 9, 27 . . . n

3^0 3^1 3^2 3^3 . . . 3^x

$3^x = n \Rightarrow \boxed{x = \log_3 n}$

As these are all loops nested

$(n - \frac{n}{3}) \times (n - \frac{n}{3}) \times (\log_3 n)$ => function complexity

$(\frac{4n^2}{9} \times \log_3 n) = \frac{4n^2 \log_3 n}{9} \Rightarrow \boxed{O(n^2 \log n)}$

* Function runs $O(n^2 \log n)$ stably. Cause of this situation function time complexity is $\boxed{\Theta(n^2 \log n)}$

③

```
define function merge(arr, l, m, r):
    set num1 to (m-l+1)
    set num2 to (r-m)
    set l_arr to [0] * (num1)
    set r_arr to [0] * (num2)

    for i in range(0, num1):
        set l_arr[i] to arr[l+i]
    for i in range(0, num2):
        set r_arr[i] to arr[m+1+i]

    set i to 0
    set j to 0
    set k to l

    while i < num1 and j < num2:
        if l_arr[i] <= r_arr[j]:
            set arr[k] to l_arr[i]
            i += 1
        else:
            set arr[k] to r_arr[j]
            j += 1

        k += 1

    while i < num1:
        set arr[k] to l_arr[i]
        i += 1
        k += 1
```

```
define furation  mergeSort(arr, l, r):

    if l < r
      set m to (l + (r-1))//2

    mergeSort(ar, l, m)
    mergeSort(ar, m+1, r)
    merge(arr, l, m, r)

define furation productPairs_helper(arr, target, pairs):
    mergeSort(ar, 0, len(arr)-1)

    set beg to 0
    set end to len(arr)-1

    while beg<end:

      set temp_t TO arr[beg]* arr[end]

      if temp_t equals target:

        Pairs.append(arr[beg], arr[end])

      if temp_t <target:

        beg+=1

      else:

        end -=1
```

```
Define function productPairs (arr, target, pairs):
    ProductPairs-helper(arr, target, pairs)

    if len(pairs) equals 0:
        Output("No pairs...")
    else:
        Output("Pairs:- " + str(pairs))
```

```
set arr to [1, 2, 3, 6, 5, 4]
set target to 6
set pairs to []

Product Pairs (arr, target, pairs)
```

* First we need to sort the array from smallest to largest. I used merge sort algorithm in this part because an nlogn solution was requested. After sorting the array, we can find pairs in linear time by performing a binary search.

Time complexity- $\theta(n \log n)$

④ Do inorder traversal of first and second tree and store on list. The complexity is the same for two operations on two trees, because trees have n-nodes.

And then merge two list to one list.

For example

Construct a new balanced tree from the merged list.
This step takes $O(2n) \approx O(n)$ times

⑤ define function isSubarray (arr1, arr2):

```
    initilize hashset

    for i in range (0, length of arr1):
        add hashset element of arr1[i]

    for i in range (0, length of arr2):
        if arr2[i] in hashset:
            continue
        else
            return False

    return True
```

* The algorithm runs linear time because adding hashset O(1) time but if "collusion" if exist then it will runs O(n) time in worst case

=> The function generally runs

O(n) linear time