

QuickSort

- The quick sort algorithm designed by Hoare is a simple yet highly efficient algorithm.
 - It works as follows:
 - Start with the given array A of n elements.
 - Consider a pivot, say $A[n]$.
 - Now, partition the elements of A into two arrays A_L and A_R such that:
 - the elements in A_L are less than $A[n]$
 - the elements in A_R are greater than $A[n]$.
 - Sort A_L and A_R , recursively.
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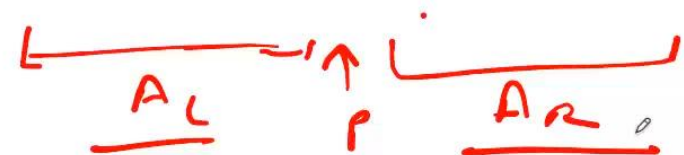
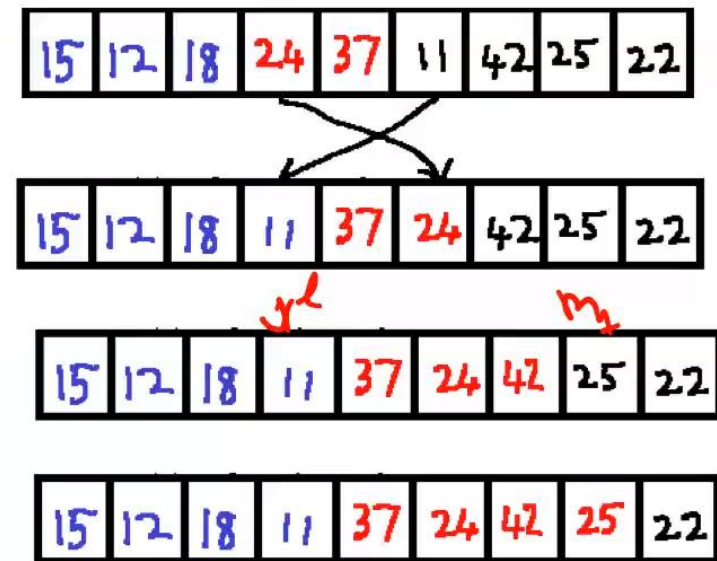
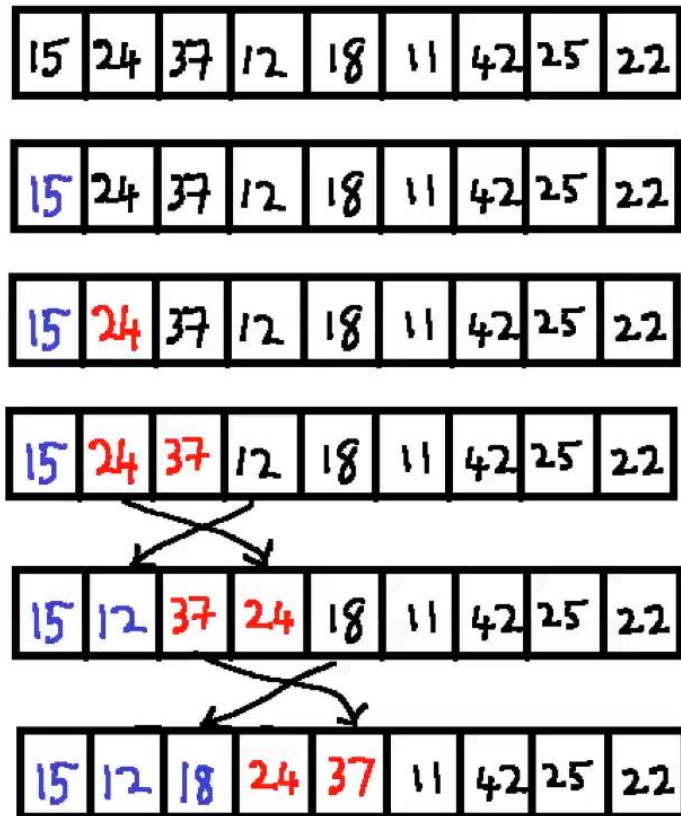
QuickSort

- How to partition ?
 - Suppose we take each element, compare it with $A[n]$ and then move it to A_L or A_R accordingly.
 - Works in $O(n)$ time.
 - Can write the program easily.
 - But, recall that space is also an resource. The above approach requires extra space for the arrays A_L and A_R
 - A better approach exists.
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QuickSort

```
Procedure Partition(A,n)
begin
  pivot = A[n];
  less = 0; more = 1;
  for more = 1 to n-1 do
    if A[more] < pivot then
      less++;
      swap(A[more], A[less]);
    end
  end
end
```

QuickSort



Analyzing Quick Sort

- We know that $|A_L| + |A_R| = n-1$.
- But, if the pivot is such that all elements are smaller (or larger) than the pivot, then $|A_L|$ (or $|A_R|$) = $n-1$.
- The recurrence relation in that case is

$$T(n) = T(n-1) + O(n).$$

- Suppose the same situation happens over every recursive call. So, the above recurrence relation holds during every recursive call.
 - When will this happen ?
-

Analyzing Quick Sort

- We know that $|A_L| + |A_R| = n-1$.
Handwritten: $n_1 + n_2 = n-1$
- But, if the pivot is such that all elements are smaller (or larger) than the pivot, then $|A_L|$ (or $|A_R|$) = $n-1$.
- The recurrence relation in that case is
$$T(n) = T(n-1) + O(n).$$
Handwritten: $T(n) = T(n-1) + \underline{T(1)} + O(n)$
- Suppose the same situation happens over every recursive call. So, the above recurrence relation holds during every recursive call.
- When will this happen ?

Analyzing Quick Sort

- In general, if the sizes of $|A_L|$ and $|A_R|$ are such that they are a constant away from each other, then the recurrence relation is:

$$T(n) = T(an) + T((1-a)n) + O(n)$$

where a is a constant < 1 .

- In practice, it turns out that most often the partitions are not too skewed.
- So, quick sort runs in $O(n \log n)$ time almost always.



Divide and Conquer

- Divide the problem P into $k \geq 2$ sub-problems P_1, P_2, \dots, P_k .
- Solve the sub-problems P_1, P_2, \dots, P_k .
- Combine the solutions of the sub-problems to arrive at a solution to P .



Divide and Conquer

- A useful paradigm with several applications.
- Examples include merge sort, convex hull, median finding, matrix multiplication, and others.
- Typically, the sub-problems are solved recursively.
 - Recurrence relation

$$T(n) = D(n) + \sum_i T(n_i) + C(n)$$

Divide time



Algorithm Merge

```
Algorithm Merge(L, R)
// L and R are two sorted arrays of size n each.
// The output is written to an array A of size 2n.
int i=1, j=1;
L[n+1] = R[n+1] = MAXINT; // so that index does not
                           // fall over
for k = 1 to 2n do
    if L[i] < R[j] then
        A[k] = L[i]; i++;
    else
        A[k] = R[j]; j++;
end-for
```

Time complexity is $O(n)$.

From Merging to Sorting

- How to use merging to finally sort?
 - Using the divide and conquer principle
 - Divide the input array into two halves.
 - Sort each of them.
 - Merge the two sub-arrays. This is indeed procedure Merge.
 - The algorithm can now be given as follows.
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Analyzing Merge Sort

- Recurrence relation for merge sort as:

$$T(n) = 2T(n/2) + O(n).$$

- This can be explained by the $O(n)$ time for merge and
 - The two subproblems obtained during the divide step each take $T(n/2)$ time.
 - Now use the general format for divide and conquer based algorithms.
 - Solving this recurrence relation is done using say the substitution method giving us $T(n) = O(n \log n)$.
 - Look at previous examples.
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Algorithm Merge Sort

```
Algorithm MergeSort(A)
begin
    mid = n/2; //divide step
    L = MergeSort(A[1..mid]);
    R = MergeSort(A[mid+1..n]);
    Merge(L, R); //combine step
end-Algorithm
```

Maxima of n Numbers

- A sequential program resembles the code below.

```
Program Maxima(A)
Begin
  int max = A[1];
  for i = 1 to n do
    if max < A[i]
      max = A[i]
    End-if
  End-for
End
```

- How do we run this program in parallel?

Parallel Merge Sort

- Need to rethink on a parallel merge algorithm
- Start from the beginning.
 - We have two sorted arrays L and R.
 - Need to merge them into a single sorted array A.
- Define the **rank** of an element x in a sorted array A as the number of elements of A that are smaller than x.
- To merge L and R, need to know the rank of every element from L and R in the merged array $L \cup R$.

Parallel Merge Sort

- Now, consider an element x in L at index k .
- How many elements of L are smaller than x ?
 - $k-1$.
- How many elements of R are smaller than x ?
 - No easy answer, but
 - can do binary search for x in R and get the answer.
 - Say k' elements in R are smaller than x .



Parallel Merge Sort

L = [8 10 12 27]

R = [15 17 24 32]

Element	8	10	12	27	15	17	24	32
Rank in L	0	1	2	3	3	3	3	4
Rank in R	0	0	0	3	0	1	2	3
Rank in L U R	0	1	2	6	3	4	5	7

L U R = [8 10 12 15 17 24 27 32]

Parallel Merge Sort

Algorithm ParallelMergeSort(A)

Begin

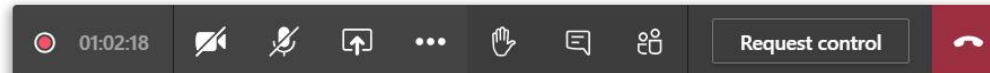
$\text{mid} = n/2$; //divide step

$L = \text{MergeSort}(A[1..\text{mid}])$;

$R = \text{MergeSort}(A[\text{mid}+1..n])$;

 ParallelMerge(L, R); //combine step

end-Algorithm



Parallel Merge Sort

- So, we just have to figure out a way to merge in parallel.
- Recall the merge algorithm as we developed it earlier.
 - Too many dependent tasks.
 - Not feasible in a parallel model.

```
for k = 1 to 2n do  
    if L[i] < R[j] then  
        A[k] = L[i]; i++;
```

```
end-for
```