### CS/IS C363 Data Structures & **Algorithms**

# **Review: Efficiency & Complexity Sorting**Time and Space Complexity

- Order Complexity and Notation
- Examples

**Cost Models** 

**Comparative Review:** 

**Sorting by Insertion and Merging Analysis** 

- A Divide-and-Conquer Algorithm
- Best and Worst cases
  - Analysis



# Linear Search Algorithm

```
// A indexed from 1 to N
index= 1;
while (index <= N) {
   if (A[index] = = x)
      return x;
}
return Not_Found;</pre>
```

### Worst case:

- Loop executes until index == N i.e. until the end of list is reached
- Size halved in each iteration N, N-1, N-2, ... 1

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Time complexity O(N)

Best Case ? Average Case ?

# Binary Search Algorithm

```
Worst case:
// A indexed from 1 to N

    Loop executes until

                                         low > high i.e. until
low = 1; high = N;
                                         size of list becomes 0
while (low <= high) {

    Size halved in each

   mid = (low + high) /2;
                                         iteration N, N/2,
  if (A[mid] = = x) return x;
                                         N/4, ... 1
  else if (A[mid] < x) low = mid +1;
                                         Number of steps is K
   else high = mid - 1;
                                                   such that 2K
                                                  i.e. logN
                                     = N
                                     steps
return Not Found;
                                     where N is input size
```

Time complexity O(logN)

# Time Complexity

- Polynomial Time Complexity
  - Time Complexity is O(Nk) for some constant k, where N is input size.
- Exponential Time Complexity
  - Time Complexity is O(2N), where N is the input size.

# Time Complexity

```
Consider the following
 algorithm:
int fact(int N) {
    j=1; prod=1;
    while (j <= N) {
        j=j+1; prod=prod*j;
    return prod;
     What is the time complexity?
     Is this polynomial time? Why
or why not?
```

# Uniform Cost vs.Logarithmic Cost

- Uniform Cost All basic operations cost same (constant) amount of time (irrespective of the data size)
- Logarithmic Cost Each operation has a cost that is proportional to the size of the data
  - Hint:

- Refer to the RAM model slide for assumptions;
- consider the call fact(100).

# **Sorting**

Comparative Review:
Sorting by Insertion and Merging
Analysis
QuickSort

- A Divide-and-Conquer Algorithm
- Best and Worst cases
  - Analysis

# Complexity of Sorting Algorithms

- Insertion Sorting:
  - ☐ Time taken for sorting N elements:
  - □ Time taken for sorting N-1 elements +
  - Time taken for inserting 1 element into sorted list
  - Time taken for inserting 1 element into sorted list:
  - $\square$  For finding the position (say the element is in Kth position): K
  - For shifting the rest of the elements: N-K
  - $\square$  Total time taken for insertion in sorted list is  $\Theta(N)$
  - Finding by binary search does it improve time complexity?

### Note on Notation

Notation: (lower bound – asymptotic lower bound)

```
\Omega(f(N)) = \{ g(n) \mid \text{there exists +ve consts.} 
c1 and n0 s.t.
```

```
c1*f(N) \le g(n) \text{ for all } n>n0
```

Notation: (tight bound – asymptotic upper and lower bound)

```
\Theta(f(N)) = \{ g(n) \mid \text{there exist } + \text{ve consts.} 
c1, c2, and n s.t.
```

```
c1*f(N) <= g(n) <= c2*f(n) for
```

# Complexity of Sorting Algorithms

Insertion Sorting – Worst case (recurrence relation):

```
 T(N) = T(N-1) + \Theta(N)  for N>1
```

$$= \Theta(1)$$
 for N=1

- Solution (techniques)
  - By substitution / iteration
- Time complexity:
  - □ Θ(N\*N)
- What about the average case?

# Complexity of Sorting Algorithms

Merge Sorting – Worst case (recurrence relation):

$$I T(N) = 2*T(N/2) + \Theta(N)$$
 for N>1

$$= \Theta(1)$$
 for N=1

- Solution (techniques)
  - By substitution / iteration
  - $\Box T(N) = \Theta(N*logN)$
  - By Master Theorem.
  - Reading Exercise: Master Theorem
- What about the average case?

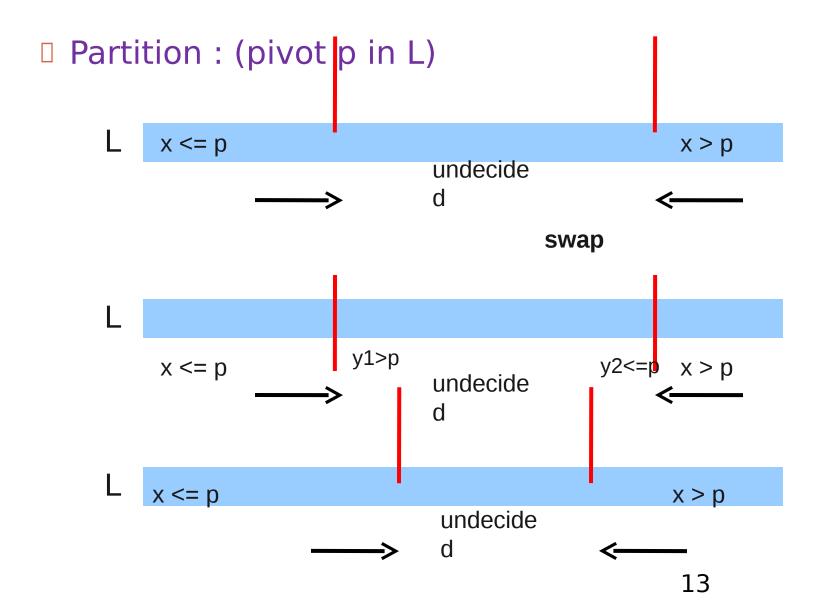
# Sorting - Comparative Review

- Insertion Sort
  - Divide & Combine: sublist of size N-1, insert
  - Time Complexity:
  - ☐ Worst case: O(N\*N)
  - Average case: O(N\*N)
  - Space Complexity:
  - □ Worst case: O(1)
  - Insertion Sort is online
  - Insertion requiresRandom Access

- Merge Sort
  - Divide & Combine: sublists of size N/2, merge
  - Time Complexity:
  - □ Worst case: O(N\*logN)
  - Average case: O(N\*logN)
  - Space Complexity:
  - Worst case: O(N)
  - Average case: O(N)
  - Merge Sort is not (fully) online Why?
  - Merge does not require Random Access

- Input: List of N elements L[0], L[1], ... L[N-1]
- Divide and Conquer:
  - Partition L into LL and LG based on a pivot p in L s.t.:

  - Combine sorted versions of LL and LG and {p}:
  - Append: LL, { p }, LG
    - Append is trivial if LL & LG were sorted in place and p is in position
    - Sorting in place requires partitioning in place:
      - Question: How do you keep **p** in between?



- Input: List of N elements L[0], L[1], ... L[N-1]
- Algorithm Schema (Divide-and-Conquer):
  - 1. Pick a pivot index j. Let p = L[j]
  - 2. Swap p out (if needed)
  - 3. Partition L based on p, into LL and LG
  - 4. Put p in place (at the right position)
  - 5. Sort LL in place
  - 6. Sort LG in place

# Complexity of Sorting Algorithms

- Complexity of Partition:
  - □ Θ(N)
- Quick Sort Time Worst case (recurrence relation):

```
 T(N) = T(N-1) + \Theta(N)  for N>1
```

$$= \Theta(1)$$
 for N=1

- $\Box$  T(N) = ?
- Quick Sort Time Best case (recurrence relation):

```
 T(N) = 2*T(N/2) + \Theta(N)  for N>1
```

$$= \Theta(1)$$
 for N=1

 $\square$  T(N) = ?

# Case Analysis

- When does the worst case occur?
- When does the best case occur?
- What is the average case complexity?

# **Sorting - QuickSort**

- Time Complexity
- Best and worst cases
  - Analyses
- Pivot Selection Median of 3, Random, Median of Medians
  - Special Cases: Small Lists, Equal-Valued Keys



# QuickSort - Time Complexity

- Time Complexity of Partition:
  - □ Θ(N) Why?
- Quick Sort Time Worst case (recurrence relation):

```
 T(N) = T(N-1) + \Theta(N)  for N>1
```

$$= \Theta(1)$$

for N=1

- $\sqcap T(N) = ?$
- When does the worst case occur?
- Quick Sort Time Best case (recurrence relation):

```
T(N) = 2*T(N/2) + \Theta(N) for N>1
```

$$=\Theta(1)$$

for N=1

$$T(N) = ?$$

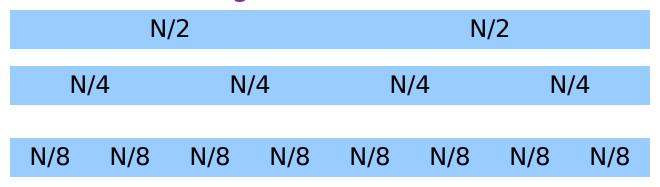
When does the best case occur?

# **QuickSort - Time Complexity**

- Average Case:
  - By assuming input distributions to be random one can compute the average case complexity.
  - Verify that O(N\*logN) is a solution to recurrence relation:
  - $\Box T(N) = \Theta(N) + (1/N) * (\sum k=1 \text{toN} (T(k-1) + T(N-k)))$
  - $\Box$  for N>1
  - T(N) = 1 for N <= 1
- But the time taken for a specific input depends heavily on the pivot(s) chosen for partitioning.

## Case Analysis

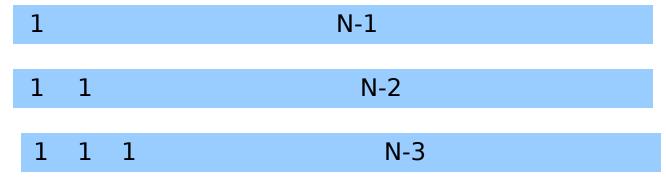
### Best case Partitioning:



- logN steps
- Partitioning work in each step is O(N)
- Time Comlexity: O(NlogN)

# Case Analysis

Worst case Partitioning:



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- N-1 steps
- Partitioning work in each step is O(N)
- Time Comlexity: O(N\*N)

# **Balanced Partitioning**

- Better time complexity with balanced partitioning
- Consider the following example scenario:

.1N	0.9N	
. 0.09 09N N	0.81N	

Time Complexity:

$$T(N) = T(N/10) + T(9N/10) + \Theta(N)$$
 $<= 2*T(9N/10) + \Theta(N)$ 
 $= O(N*log10/9 (N))$ 

### **Pivot Selection**

- Biased pivots result in unbalanced partition
  - □ Pivot p such that p < x for most x in L</p>
  - $\square$  Symmetrically, p > x for most x in L
- Static/Fixed techniques for pivot selection repeat the bias at every level
  - E.g. First element of the list as a pivot in a (mostly) sorted list
- De-biasing Solution:
  - Adaptive selection of pivots
  - Typically done by sampling the input

# Pivot Selection Techniques

- Median of 3:
- Median of Medians:
- QuickSelect (selecting Kth smallest element)

# Pivot Selection Techniques

### Median of 3:

- Median of first, middle, and last element in the (sub)list
- Exclude these elements from partitioning process

### Median of Medians:

- For every 5 contiguous elements find the median by direct comparison ==> N/5 medians
- Obtain the median of these N/5 medians
- How? Sort? QuickSelect?
- QuickSelect (selecting Kth smallest element)

### Pivot Selection - QuickSelect

- Median of Medians:
  - □ For every 5 contiguous elements find the median by direct comparison ==> N/5 medians
  - Obtain the median of these N/5 medians
  - How? Sort? QuickSelect?
- QuickSelect (selecting Kth smallest element)
  - Partition the list by a pivot p;
  - p is in its correct position, say J
  - 3. if K = J done,

Q: What is the time complexity of QuickSelect?

if K<J select Kth smallest from left sub list
if K>J select K-J th smallest element from right sub
list

### **Pivot Selection**

- Randomized QuickSort
  - Select pivot index uniformly randomly between first and last (indices).
  - Need a (good) random number generator
- Is there a random number?
  - Random sources
  - Bit Selection
  - Repeated coin toss
  - Cost of random number generation
  - Pseudo-random number generators

### **Small Lists**

- Insertion Sort performs better than QuickSort on small lists.
  - Why?
  - So, what?
- Combine the two!
  - Invoke Insertion Sort inside QuickSort when size of the list is small.
  - Alternatively, ignore, small sized lists inside QuickSort, and do an insertion Sort (on the full list)
  - Question: Why does this work (efficiently)?
  - Time Complexity (expected):
  - O(k\*k\*N + N\*log(N)) where k is the threshold (below which InsertionSort performs better than QuickSort)

# **Equal Values**

- QuickSort performs badly when the same key occurs multiple times
- Solution: 3-way partition
  - Maintain an additional partition for elements equal to the pivot
  - Exercise: Implement this!

