

# Introduction and Motivation

# Design and Analysis of Algorithms

- Design means designing an algorithm to solve a problem
- Analysis means measuring efficiency of the algorithm in terms of the input size
- We measure in two ways:
  - **Time** – Relationship between time taken to run the algorithm and input size
  - **Space** – Relationship between space consumed and input size

# Measuring Running Time

- As space is easily available
- We shall concentrate of the time taken by the algorithms
- Analysis is independent of underlying hardware
  - We don't use actual time
  - Measure in terms of "basic operations"

# Measuring Running Time

- Typical basic operations
  - Compare two values
  - assign a value to a variable
  - Swap two numbers
    - `tmp ← x`
    - `x ← y`
    - `y ← tmp`
- Three assignments constants can be ignored

# Measuring Running Time

- Running time depends on input size
- Longer array will take more to sort than shorter arrays
- But input of same size may take different amount of time

# Measuring Running Time

## Linear Search

Search 'K' in an unsorted array A

$i \leftarrow 0$

while  $i < n$  and  $A[i] = k$  do

$i \leftarrow i + 1$

if  $i < n$  return  $i$

else return  $-1$

# Measuring Running Time

- Complexity of Linear Search
- Input 2 – 34, 12, 36, 78, 52 search for 34 – Best
- Input 3 – 34, 12, 36, 78, 52 search for 70 – Worst
- Average case may be computed for better idea
- But not possible to calculate for all cases
- Need probability distribution over inputs
- Good upper bound is got from worst case

# How Important is Complexity?

## Sorting Algorithms

Naive –  $O(n^2)$

Best –  $O(n \log n)$

Typical CPU process upto  $10^8$  operations per second

Moore's law – speed is saturated

Useful for approximate calculation



# Computing Time for Sorting Mobile Users

- Sort a list of mobile numbers in India
- Roughly around  $10^9$  numbers
- Naive algorithm ( $n^2$ ) need  $10^{18}$  operations
- $10^8$  operations per second so it needs  $10^{10}$  seconds
- 2778000 hours
- 115700 days
- 300 years !

# Computing Time for Sorting Mobile Users

- Smart algorithm ( $n \log n$ )
- need  $3 \times 10^{10}$  operations
- about 300 seconds
- 5 minutes

# Computing Time for a Video Game

- Several objects in the screen
- Basic step – find closest pair of objects
- Given 'n' objects, naive algorithm takes  $n^2$  time
  - For each pair of objects, compute their distance
  - Report minimum distance over all such pairs
- There is a clever algorithm that takes  $n \log n$  time

# Computing Time for a Video Game

- High resolution monitor has 2500 X 1500 pixels
  - 3.75 million points
- suppose we have  $500,000 = 5 \times 10^5$  objects
- Naive algorithm will take  $25 \times 10^{10}$  steps = 2500 secs
- 42 minutes to respond
- Smart  $n \log n$  algorithm takes a fraction of a second

# Typical Functions T(n)

Input	log n	n	n log n	n <sup>2</sup>	n <sup>3</sup>	2 <sup>n</sup>	n!
10	3.3	10	33	100	1000	1000	10 <sup>6</sup>
100	6.6	100	66	10 <sup>4</sup>	10 <sup>6</sup>	10 <sup>30</sup>	10 <sup>157</sup>
1000	10	1000	10 <sup>4</sup>	10 <sup>6</sup>	10 <sup>9</sup>		
10 <sup>4</sup>	13	10 <sup>4</sup>	10 <sup>5</sup>	10 <sup>8</sup>	10 <sup>12</sup>		
10 <sup>5</sup>	17	10 <sup>5</sup>	10 <sup>6</sup>	10 <sup>10</sup>			
10 <sup>6</sup>	20	10 <sup>6</sup>	10 <sup>7</sup>				
10 <sup>7</sup>	23	10 <sup>7</sup>	10 <sup>8</sup>				
10 <sup>8</sup>	27	10 <sup>8</sup>	10 <sup>9</sup>				
10 <sup>9</sup>	30	10 <sup>9</sup>	10 <sup>10</sup>				
10 <sup>10</sup>	33	10 <sup>10</sup>					

10<sup>9</sup> operations is considered to be fine

– 10 secs

10<sup>10</sup> is too long

# Quantifying Efficiency

- Big O – Upper bound
- Omega – Lower bound
- Theta – Tight bound

# Insertion sort

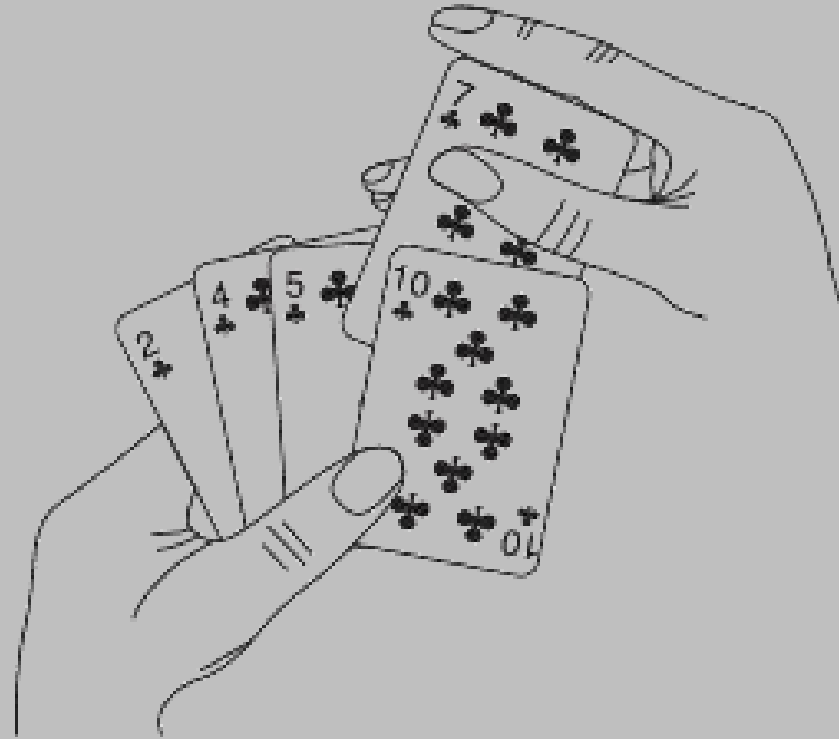
- **Input:** A sequence of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$ .
- **Output:** A permutation (reordering)  $\langle a'_1, a'_2, \dots, a'_n \rangle$  of the input sequence such that  $a'_1 \leq a'_2 \leq \dots \leq a'_n$
- Numbers that we wish to sort are also known as the keys
- Although conceptually we are sorting a sequence, the input comes to us in the form of an array with  $n$  elements.

# Insertion sort

- Works the way many people sort a hand of playing cards
- Start with an empty left hand and the cards face down on the table
- then remove one card at a time from the table and insert it into the correct position in the left hand
- To find correct position for a card, we compare it with each of the cards already in the hand, from right to left



# Insertion sort



# Insertion sort

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



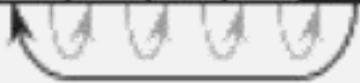
1	2	3	4	5	6
2	5	4	6	1	3



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



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



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



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

# Insertion sort

INSERTION-SORT( $A$ )

```
1  for  $j = 2$  to  $A.length$ 
2       $key = A[j]$ 
3      // Insert  $A[j]$  into the sorted sequence  $A[1..j-1]$ .
4       $i = j - 1$ 
5      while  $i > 0$  and  $A[i] > key$ 
6           $A[i + 1] = A[i]$ 
7           $i = i - 1$ 
8       $A[i + 1] = key$ 
```

- $j$  indicates “current element” being inserted into sorted list
- At the beginning of each iteration , subarray consisting of elements  $A[1..j-1]$  constitutes sorted elements, and remaining subarray  $A[j+1..n]$  corresponds to elements yet to be sorted

# Correctness of Insertion sort

- Identifying Loop Invariant
- Showing that the loop invariant is true in the Initialization, Maintenance and Termination

# Correctness of Insertion sort

- Identifying Loop Invariant
- **Fact:**  $A[1, \dots, j-1]$  are the elements originally in positions 1 through  $j-1$ , but now in sorted order.
- At the start of each iteration of the for loop of lines 1 – 8, the subarray  $A[1, \dots, j-1]$  consists of the elements originally in  $A[1, \dots, j-1]$ , but in sorted order.

# Correctness of Insertion sort

- **Initialization:** It is true prior to the first iteration of the loop
- **Maintenance:** If it is true before an iteration of the loop, it remains true before the next iteration
- **Termination:** When the loop terminates, the invariant gives us a useful property that helps show that the algorithm is correct

# Correctness of Insertion sort

- We use first two properties similar to mathematical induction
- First condition is similar to base case and the second condition is something similar to inductive step
- We use loop invariant along with the condition that caused the loop to terminate.
- Induction goes infinitely

# Correctness of Insertion sort – Initialization

- When  $j = 2$ , we have a subarray of size 1, an array of size is sorted by itself



## Correctness of Insertion sort – Maintenance

- Body of for loop works by moving  $A[j-1]$ ,  $A[j-2]$ ,  $A[j-3]$  , and so on by one position to right until it finds proper position for  $A[j]$  (lines 4 – 7), at which point it inserts value of  $A[j]$  (line 8)
- Subarray  $A[1 .. j]$  then consists of elements originally in  $A[1 .. j]$  , but in sorted order
- Incrementing  $j$  for next iteration of for loop then preserves the loop invariant.

# Correctness of Insertion sort – Termination

- What happens when the loop terminates
- Condition causing for loop to terminate is that  $j > A.length$
- Because each loop iteration increases  $j$  by 1, we must have  $j = n+1$  at that time
- Substituting  $n+1$  for  $j$  in the wording of loop invariant, we have that the subarray  $A[1..n]$  consists of the elements originally in  $A[1..n]$ , but in sorted order
- Hence the entire array is sorted.

# Frequency Count Method

- We count number of times one instruction is executing

```
Algorithm Sum(A, n)
```

```
{
```

```
    s = 0;
```

```
    for(i=0; i<n; i++)
```

```
    |    s = s+A[i];
```

```
    return s;
```

```
}
```

1

$1 + (n+1) + n$

n

$f(n) = 2 + (n+1) + 2*n$

$f(n) = 3 + 3*n$

$f(n) = O(g(n)) \Rightarrow f(n) \leq c*n$  for some c and  $n_0$

$f(n) = O(n)$

# Frequency Count Method

Algorithm Sum1(A,n)	
{	1
s = 0;	
for(i=1; i<n; i=i+2)	$1 + (n+1)/2 + n/2$
s = s+A[i];	$n/2$
return s;	
}	$f(n) = 1.5 + n$

$f(n) = O(g(n)) \Rightarrow f(n) \leq c \cdot n$  for some  $c$  and  $n_0$

$f(n) = O(n)$

# Frequency Count Method

```
Algorithm nested(A,n)
```

```
{
```

```
    s = 0;
```

```
    for(i=0; i<n; i++)
```

```
        for(j=0; j<n; j++)
```

```
            s = s+A[i]+A[j];
```

```
    return s;
```

```
}
```

1

n+1

n\*(n+1)

n\*n

$f(n) = 2*n^2 + n + 1$

$f(n) = O(g(n)) \Rightarrow f(n) \leq c*n^2$  for some c and n0

$f(n) = O(n^2)$

# Frequency Count Method

```
Algorithm nested(A,n)
{
    s = 0;
    for(i=0;i<n;i++)
        for(j=0;j<i;j++)
            s = s+A[i]+A[j];
    return s;
}
```

1

$n+1$

$1+2+3+4+....$

$n*(n+1)/2$

$f(n) = n^2/2 + n / 2$

$f(n) = O(g(n)) \Rightarrow f(n) \leq c*n^2$  for some  $c$  and  $n_0$

$f(n) = O(n^2)$

# Insertion sort

INSERTION-SORT( $A$ )

```
1  for  $j = 2$  to  $A.length$ 
2       $key = A[j]$ 
3      // Insert  $A[j]$  into the sorted sequence  $A[1..j-1]$ .
4       $i = j - 1$ 
5      while  $i > 0$  and  $A[i] > key$ 
6           $A[i + 1] = A[i]$ 
7           $i = i - 1$ 
8       $A[i + 1] = key$ 
```