#### Overview

- MOTIVATION
- ALGO DEF.
- ANALYSIS Seq. Search
- TIME
  WORK DONE
  WORST, AV CASE

**SPACE** 

- OPTIMALITY
- ORDER NOTATION
- Design Bin Search.

#### • MOTIVATION

- Algorithms: How to do things on a computer?
- Other application areas study specialized algorithms.

e.g.

Operating Systems

Compiler Design

Artificial Intelligence

Data Base Management Systems

- Algorithm Design Techniques
   Divide and conquer, greedy, Dynamic Programming etc.
- Basic Algorithms
   sorting, graph algo, matrix multiplication, NP Complete problems, parallel algos.

#### • HISTORY

Phrase Algorithm; Persian Author

- Abu Jafar Mohammed ibn Musa *al Khowarizmi* (825 AD)
- wrote a Math textbook
- al Khowarizmi: from the town of Khowarazm (now Khiva, Uzbekistan)

#### • ALGORITHM DEFINITION

An algorithm is composed of a finite number of steps, each of which may require one or more operations.

- each OPERATION executable on a computer.
- compute 5/0 is not WELL DEFINED.
- algorithms must TERMINATE after a finite number of operations.
- PROCEDURE: a nonterminating algorithm. e.g. OS.

#### **DESIGN & ANALYSIS**

- GOAL Distinguish Algorithms.
- TIME Computer Dependent. AMOUNT OF WORK - COST
  - proportional to the number of **basic operations**.
  - computer independent.

e.g. matrix multiplication basic operation - multiplication & addition.

# Example Problem: Sequential Search

**Problem:** Given an array L[1..n], containing n DISTINCT entries, find the index of x, if  $x \in L$ , else return 0.

**Input:** array L, array size n, search item x.

**Output:** If  $x \in L$  then index of x in L, else 0.

## Algorithm:

Abstract Level Description: Scan L left to right looking for x in L & return index.

#### Pseudocode

- 1. index  $\leftarrow$  1
- 2. While index  $\leq n$
- 3. **if** L[index] = x
- 4. return (index)
- 5. else
- 6.  $index \leftarrow index + 1$
- 7. return (0)

#### Work done

Steps 1. assignment to a register variable

While loop, if 
$$x = L[k]$$

- 2. k comparisons of register variables
- 3. k comparisons of x with L entry.
- 6. k-1 increments of register variable.

Since step 3 is costliest & other steps are executed no more times, basic operation is comparison of a memory and a register variable.

$$cost = k, & if x = L[k] \\
cost = n, & if x \notin L$$

# Time Complexity

### Best Case time

- k = 1, x = L[1], best-case input.
- B(n) = 1, as a function of input size.

### Worst Case time

- x = L(n) or  $x \notin L$
- W(n) = n

## Average-Case cost?

- Let  $x \in L$
- Further, let x is equally likely to be in any position 1 through n.
- $E_k$ : event that x = L[k]
- Probability of  $E_k$  happening

$$P(E_k) = 1/n, \quad 1 \le k \le n$$
  
 $t(E_k) = \text{cost or time when } L[k] = x$   
 $= k$ 

Thus, average cost

$$A(n) = P(E_1)t(E_1) + P(E_2)t(E_2) + \dots + P(E_n)t(E_n)$$

$$= \sum_{k=1}^{n} P(E_k)t(E_k)$$

$$= \sum_{k=1}^{n} \frac{1}{n}k$$

$$= \frac{1}{n} \sum_{k=1}^{n} k$$

$$= \frac{1}{n} (\frac{n(n+1)}{2})$$

$$= \frac{n+1}{2}$$

### Did we average over all possible inputs?

Let us allow the possibility that x might not be in L.

$$E_0 = \text{event that } x \not\in L$$

Let

$$P(x \in L) = q$$
  
 
$$P(x \notin L) = 1 - q = P(E_0)$$

For  $k \geq 1$ ,

$$P(E_k) = P(x \in L \text{ and } x = L[k])$$
  
=  $P(x \in L)P(x = L[k] \text{ given that } x \in L)$   
=  $q\frac{1}{n} = \frac{q}{n}$ 

Thus,

$$A(n) = \sum_{k=0}^{n} P(E_k)t(E_k)$$

$$= \sum_{k=1}^{n} P(E_k)t(E_k) + P(E_0)t(E_0)$$

$$= \sum_{k=1}^{n} \frac{q}{n}k + (1-q)n$$

$$= \frac{q}{n}\sum_{k=1}^{n} k + (1-q)n$$

$$= \frac{q}{n}\frac{n(n+1)}{2} + (1-q)n$$

$$= q\frac{n+1}{2} + (1-q)n$$

Thus,

• if 
$$q = 1$$
,  $x \in L$ ,  $A(n) = \frac{n+1}{2}$ 

• if 
$$q = 0$$
,  $x \notin L$ ,  $A(n) = n$ 

• if q = 1/2,

$$A(n) = \frac{1}{2} \frac{n+1}{2} + \frac{1}{2}n$$

$$= \frac{n+1+2n}{4}$$

$$= \frac{3n}{4} + \frac{1}{4}$$

Cost, Work = time complexity.

# Space Complexity

Extra space used apart from the input and the program (and the output, if required by specification).

$$S(n) = 1$$
 (1 register variable)

# Trade-off between time & space

If number are between 1 and 100, then a prepared index  $\operatorname{array} A[1\dots 100]$  such that

$$A[i] = j$$
, if  $L[j] = i$ ; else  $A[i] = 0$ 

can yield W(n) = O(1).

 $(characteristic\ array)$ 

## **Optimality**

- Is the sequential search algo. optimal?
- Is there another algo. which solves the same problem using fewer number of comparisons?
- Theorem: In the worst case,

$$W(n) = n$$

- Proof by contradiction
  - If there is another algo B whose W(n) < n, then the element not seen by B may be x & B would be incorrect.
    - $(O(\log n))$  on an ordered list & O(1) in a characteristic vector alternatives only when performing multiple searches.)
  - **Lower Bound** on the number of comparisons for searching in an unordered list is n.

# Problem Complexity

<u>Def.</u> (Worst case) Complexity of a problem is the min number of operations needed to solve the problem in the worst case using particular resources.

eg. Matrix Multiplication  $C_{nn} = A_{\text{n x n}} * B_{\text{n x n}}$   $C_{ij} = A_{i1}B_{1j} + A_{i2}B_{2j} + \cdots + A_{in}B_{nj}$ 

for 
$$i \leftarrow 1$$
 to  $n$   
for  $j \leftarrow 1$  to  $n$   
 $C_{ij} \leftarrow 0$   
for  $k \leftarrow 1$  to  $n$   
 $C_{ij} = C_{ij} + A_{ik}B_{kj}$ 

- Basic operation: Multiplication (addition can be ignored)
- $W(n) = n^3$
- lower bound on number of multiplications =  $n^2$  lower bound on the problem complexity of matrix multiplication =  $n^2$ .
- best algorithm found has complexity  $n^{2.367}$

# Average-case problem complexity

Space complexity of a problem  $\bullet$  matrix multiplication O(1)

- sorting on comparison model O(1).
- matching parentheses O(n).