

# Decrease-and-Conquer

## *CS3230: Design and Analysis of Algorithms*

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# Chapter 5: Decrease-and-Conquer

## Topics: What we will cover today

- Insertion sort
- Depth-first search and Breadth-first search
- Topological sorting
- Algorithms for generating combinatorial objects
- Decrease-by-a-constant-factor algorithms
- Variable-size-decrease algorithms

# The Decrease-and-Conquer Design Strategy

- The **decrease-and-conquer** technique exploits the relationship between a solution to a problem instance and a solution to a smaller instance of the same problem.
- The relationship may be exploited top-down with recursion or bottom-up with iteration. We may refer to the latter as **increase-and-conquer**.
- The three major variations are:
  - decrease by a constant (often 1),
  - decrease by a constant factor (often half), and
  - variable size decrease.

# Insertion Sort

- To sort the keys given in the array  $A[0..n-1]$ , we first sort the prefix  $A[0..n-2]$ , and then insert  $A[n-1]$  into  $A[0..n-2]$  at the right place.
- There are three ways to find the right place
  1. scan from  $A[0]$  to  $A[n-2]$  until the first  $A[i]$  with  $A[i] \geq A[n-1]$  or the prefix is exhausted;
  2. scan from  $A[n-2]$  to  $A[0]$  until the first  $A[i] \leq A[n-1]$  or the prefix is exhausted;
  3. use binary search.

# An Insertion Sort Algorithm

```
INSERTIONSORT (  $A[0..n - 1]$  )  
  // Input:  $n$  keys in  $A[0..n - 1]$   
  for  $i \leftarrow 1$  to  $n - 1$  do  
    // Insert  $A[i]$  to  $A[0..i - 1]$   
     $v \leftarrow A[i]$   
     $j \leftarrow i - 1$   
    while  $j \geq 0$  and  $A[j] > v$  do  
       $A[j + 1] \leftarrow A[j]$   
       $j \leftarrow j - 1$   
    od  
     $A[j + 1] \leftarrow v$   
  od
```

## *Insertion Sort: An Example*

89		45	68	90	29	34	17
45	89		68	90	29	34	17
45	68	89		90	29	34	17
45	68	89	90		29	34	17
29	45	68	89	90		34	17
29	34	45	68	89	90		17
17	29	34	45	68	89	90	

# An Analysis of the Insertion Sort Algorithm

- Let  $C(n)$  be the key comparison count.
- For the worst case (i.e., the input is already sorted reversely):

$$C(n) = \sum_{i=1}^{n-1} \sum_{j=0}^{i-1} 1 = \frac{(n-1)n}{2} \in \Theta(n^2).$$

- For the best case (i.e., the input is already sorted):

$$C(n) = \sum_{i=1}^{n-1} 1 = n - 1 \in \Theta(n).$$

# An Analysis of the Insertion Sort Algorithm

- For the average case, it is known that

$$C(n) \approx \frac{n^2}{4} \in \Theta(n^2).$$

- Space efficiency: in-place ( $O(1)$ )
- Stability: yes
- Variation: binary insertion sort



## *High Level "Visualizing" of Sorting Algorithms*

- To understand the differences between the different sorting algorithms we can divide the data array into a “**prefix**” and a “**suffix**”.
- Then we can describe at a high level how a sorting algorithm does its work on the prefix and the suffix.

## "Visualizing" Selection Sort

prefix  $\leftarrow$  null;    suffix  $\leftarrow A[0..n - 1]$

while suffix  $\neq$  null do

    swap the first element of suffix with  
    a minimum element of suffix;

    make the first element of suffix to become  
    the last element of prefix

od

## "Visualizing" Bubble Sort

prefix  $\leftarrow A[0..n - 1]$ ;    suffix  $\leftarrow$  null

while prefix  $\neq$  null do

    sort every pair of adjacent prefix elements  
    from left to right;

    make the last element of prefix to become  
    the first element of suffix

od

## "Visualizing" Insertion Sort

prefix  $\leftarrow A[0]$ ;    suffix  $\leftarrow A[1..n - 1]$

while suffix  $\neq$  null do

    remove the first element of suffix and

    insert it into the prefix

od

# Graph Algorithms

- Graph notations and terms:
  - Set of vertices  $V$ , number of vertices  $|V|$ .
  - Set of edges  $E$ , number of edges  $|E|$ .
  - Graph  $G = (V, E)$ .
  - Types of graphs: **undirected**, **directed**, and **weighted**.
  - Representations: **adjacency matrix** and **adjacency lists**.
  - **Tree edge**: direct edge between a parent and a child vertex.
  - **Back edge** in DFS: connects a vertex to an ancestor, other than the parent.
  - **Cross edge** in BFS: connects a vertex to a visited vertex other than its immediate predecessor.

# Graph Traversal

- Many problems require processing of all graph vertices (and edges in a systematic fashion.
- Graph traversal algorithms:
  - There are two systematic ways to visit all the vertices of a graph: **depth-first search** (DFS) and **breadth-first search** (BFS).

# Depth-First Search

- For each connected component, visit a given vertex. On the next search, visit an unvisited vertex adjacent to the vertex it has just visited; if there is no such vertex, back up to where it came from and re-start the search.
- Uses a stack:
  - A vertex is pushed onto the stack when it is reached for the first time.
  - A vertex is popped off the stack when it becomes a dead end, i.e., when there is no adjacent, unvisited vertex.
- “Redraws” graph in a tree-like fashion (with tree edges and back edges for undirected graph).

# A Depth-First Search Algorithm

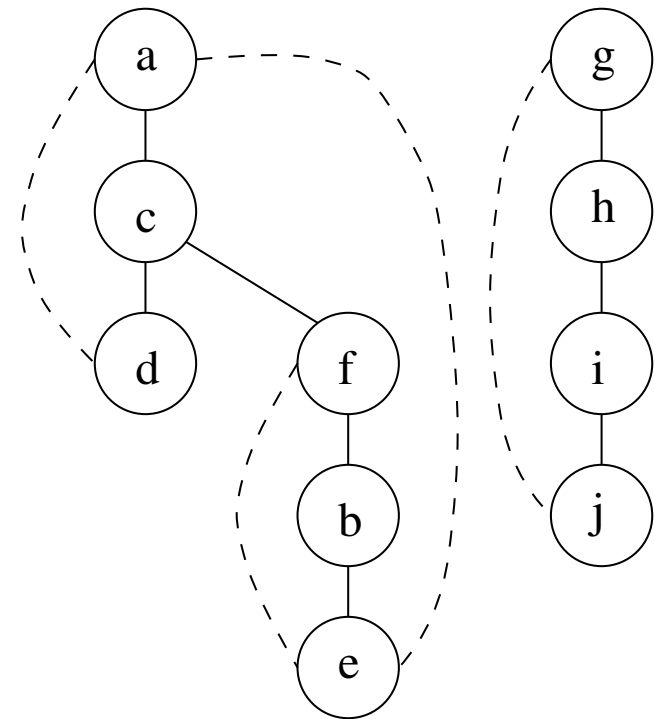
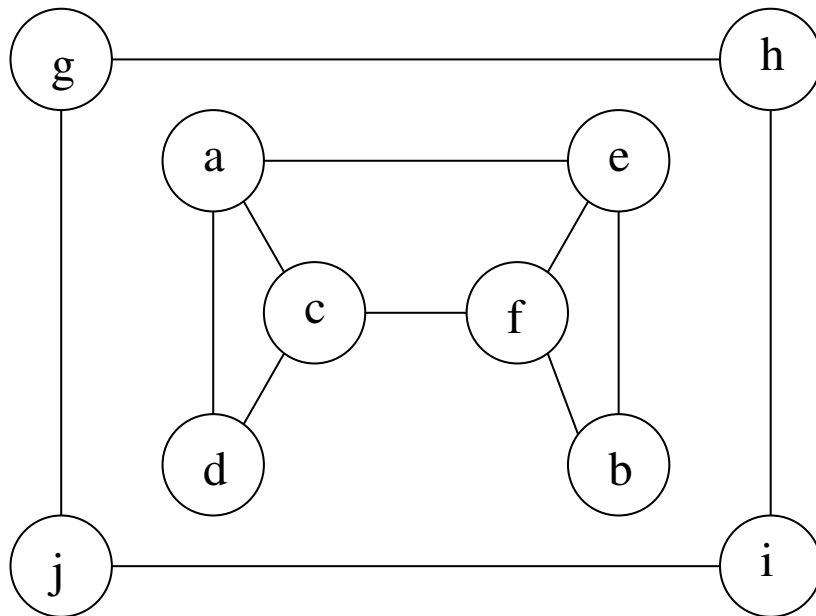
```
DFS(  $G$  )           //  $G = (V, E)$ ,  $|V| = n$ 
    for each vertex  $v$  in  $V$  mark  $v$  as unvisited    // Initialization
    for each vertex  $v$  in  $V$ 
        if  $v$  is unvisited then  $dfs( v )$ 
    end
```

```
 $dfs( v )$ 
    mark  $v$  as visited
    for each vertex  $w$  adjacent to  $v$ 
        if  $w$  is unvisited then  $dfs( w )$ 
    end
```



# A Depth-First Search Example

- Dashed lines: back edges.



# Notes on Depth-First Search

- DFS can be implemented with graphs represented as adjacency matrices and adjacency lists.
- It yields two distinct orderings of vertices.
  - Order in which vertices are first encountered (pushed onto stack).
  - Order in which vertices become dead-ends (popped off stack).
- Applications
  - Checking connectivity, finding connected components
  - Checking acyclicity
  - Searching state-space of problems for solution (AI)

# *An Analysis of the Depth-First Search Algorithm*

- The cost is clearly proportional to the size of the data structures used to represent the graph.
- For the adjacency matrix representation, every adjacency row has  $|V|$  entries. Thus the traversal time is in  $\Theta(|V|^2)$ .
- For the adjacency lists representation, the total number of adjacency lists entries is  $2|E|$ . Thus the traversal time is  $\Theta(|V| + |E|)$ .

## *Breadth First Search*

- While depth first search tries to go as far as possible, breadth first search tries to stay as near as possible.
- It first visits the start vertex.
- It then visits all the vertices 1-edge away, then visits all the vertices 2-edge away, etc., until all the vertices in the component have been visited.
- While depth first search is greatly facilitated by the function call stack, breadth first is greatly facilitated by a queue.

## A Breadth-First Search Algorithm (Initialization)

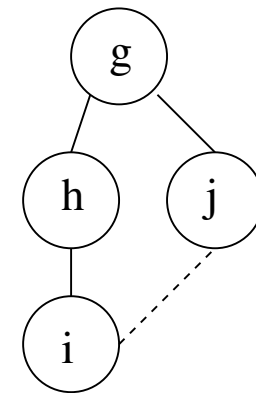
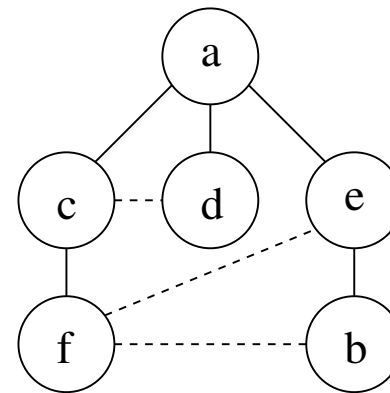
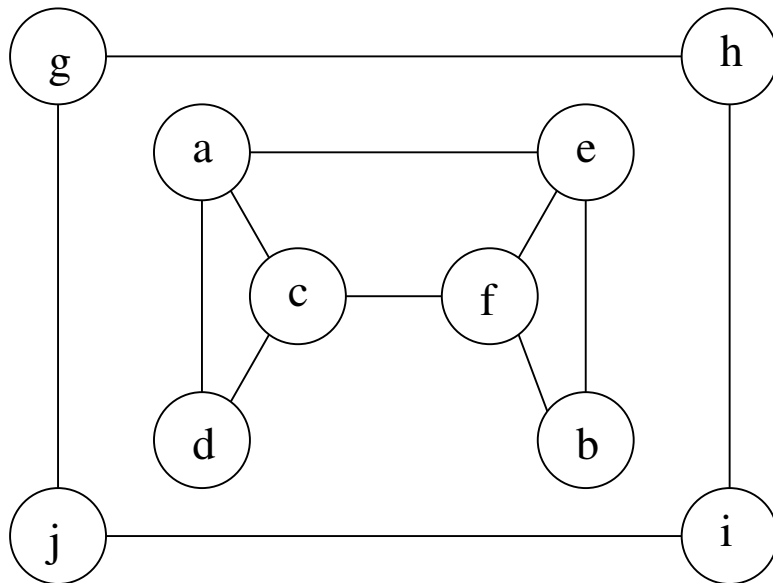
```
BFS(  $G$  )           //  $G = (V, E)$ 
  for each vertex  $v$  in  $V$  mark  $v$  as unvisited    // Initialization
  for each vertex  $v$  in  $V$ 
    if  $v$  is unvisited then  $bfs( v )$ 
end
```

## A Breadth-First Search Algorithm (Iteration)

```
bfs( v )  
  mark v as visited; init_queue(); enqueue( v )  
  while queue  $\neq$  empty do  
    u  $\leftarrow$  dequeue()  
    for each vertex w adjacent to u do  
      if w is unvisited then  
        mark w as visited; enqueue( w )  
      fi  
    od  
  od  
end
```

# A Breadth-First Search Example

- Dashed lines: cross edges.



# An Analysis of the Breadth-First Search Algorithm

- Similar to depth first search, breadth-first search also has to check all the adjacent vertices of each vertex.
- The all cases time efficiency is also  $\Theta(|V|^2)$  with adjacency matrices and  $\Theta(|V| + |E|)$  with adjacency lists.
- It yields a single ordering of vertices (i.e., order added/deleted from queue is the same).
- Applications: same as DFS; but it can also find paths from a vertex to all other vertices with the smallest number of edges.



# Topological Sorting

- A directed graph without cycles is a directed acyclic graph (**dag**).
- The vertices of a dag can be sorted topologically.
- A permutation of the  $n$  vertices  $v_1, \dots, v_n$  of the dag  $G = (V, E)$

$$v_{i_1}, \dots, v_{i_n}$$

is a **topological sort** of the dag if  $(v_{i_p}, v_{i_q})$  is not an edge whenever  $p > q$ .

- That is,

$$E \subseteq \{(v_{i_p}, v_{i_q}) : p < q\}.$$

## Topological Sorting by Depth-First-Search

- Observe that when the function  $\text{dfs}(v)$  returns, all vertices that can be reached from  $v$  must have been visited.
- Thus the reverse order of the return of  $\text{dfs}(v)$  gives a topological sort.

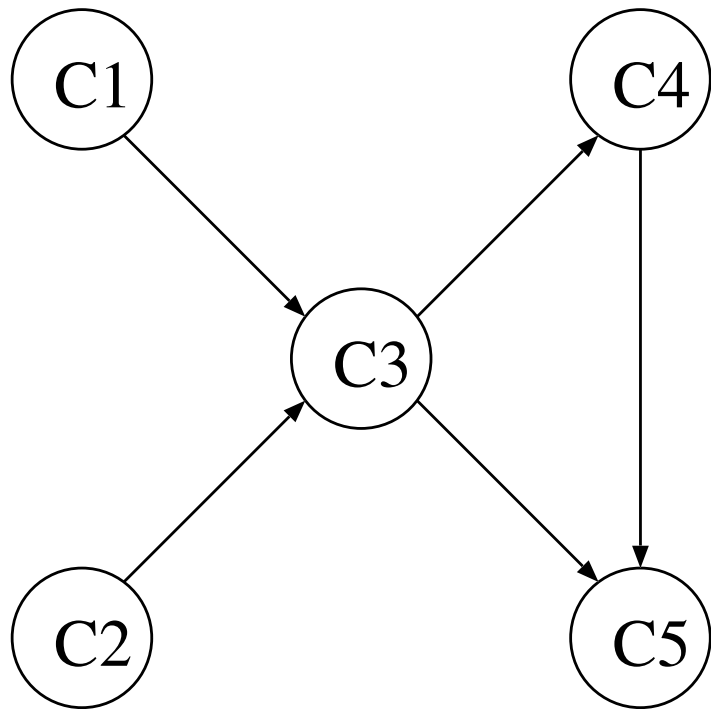
# A Topological Sorting Example

- Example:

A student needs to take 5 required courses:  $\{C1, C2, C3, C4, C5\}$ . The following pre-requisites need to be met:  $C1$  and  $C2$  have none;  $C3$  requires  $C1$  and  $C2$ ;  $C4$  requires  $C3$ ; and  $C5$  requires  $C3$  and  $C4$ .

The student can only take one course per semester.

# A Topological Sorting Example



The popping-off order:

C5, C4, C3, C1, C2

The topologically sorted list:

C2    C1 → C3 → C4 → C5

# A Topological Sort Algorithm by Depth-First-Search

```
// Produce a reverse topological sort
TS_DFS(  $G$  )           //  $G = (V, E)$ 
    for each vertex  $v$  in  $V$  mark  $v$  as unvisited
    for each vertex  $v$  in  $V$ 
        if  $v$  is unvisited then  $dfs( v )$ 
    end

 $dfs( v )$ 
    mark  $v$  as visited
    for each vertex  $w$  adjacent to  $v$ 
        if  $w$  is unvisited then  $dfs( w )$ 
    print  $v$ 
end
```

# Algorithms for Generating Combinatorial Objects

- Given a set of  $n$  objects, the most important types of combinatorial objects are
  - the  $n!$  permutations,
  - the  $\binom{n}{k}$   $k$ -combinations,
  - the  $2^n$  subsets.
- Note that

$$\binom{n}{0} + \cdots + \binom{n}{n} = (1 + 1)^n = 2^n.$$

# Generating Permutations

- Consider the problem of finding the  $n!$  permutations of the sequence  $1\ 2 \cdots n$ .
- There are at least three methods: decrease by 1 (actually should be increase by 1), the Johnson-Trotter algorithm, and the lexicographic order method.
- The increase-by-1 algorithm enjoys the minimal-change property: each permutation can be obtained from its immediate predecessor by exchanging two adjacent digits.

## *An Illustration of the Increase-By-1 Algorithm: 3!*

- Generate 1.
- Insert 2 to 1 from right to left: 12, 21.
- Insert 3 to 12 from right to left: 123, 132, 312.
- Insert 3 to 21 from left to right: 321, 231, 213.
- Observe the minimal-change property of the permutations

123, 132, 312, 321, 231, 213.



## *An Illustration of the Increase-By-1 Algorithm: 4!*

- Insert 4 to 123 from right to left: 1234, 1243, 1423, 4123.
- Insert 4 to 132 from left to right: 4132, 1432, 1342, 1324.
- Insert 4 to 312 from right to left: 3124, 3142, 3412, 4312.
- Insert 4 to 321 from left to right: 4321, 3421, 3241, 3214.
- Insert 4 to 231 from right to left: 2314, 2341, 2431, 4231.
- Insert 4 to 213 from left to right: 4213, 2413, 2143, 2134.

•  
•  
•

## *Observe Minimal-Change Property of the Permutations*

1234, 1243, 1423, 4123, 4132, 1432, 1342, 1324, 3214, 3241, 3412, 4312,  
4321, 3421, 3241, 3214, 2314, 2341, 2431, 4231, 4213, 2413, 2143, 2134.

## *Mobile Digits of the Johnson-Trotter Algorithm*

- Each element  $e$  in a permutation is given a left-to-right direction  $e.$  or a right-to-left direction  $.e$ .
- An element in a permutation is mobile if its direction points at a smaller element.
- For example, in the permutation 3..24..1, the digits 3 and 4 are mobile, the digits 1 and 2 are not.

# The Johnson-Trotter Permutation Generation Algorithm

```
JOHNSONTROTTER(  $n$  )           //  $n$  is a positive integer
  initialize the first permutation with  $.1.2 \dots .n$ 
  while previous permutation has a mobile element do
    find the largest mobile element  $k$ 
    swap  $k$  with the element it is pointing to
    reverse direction of all elements larger than  $k$ 
    add the new permutation to the list
  od
end
```

## *The Johnson-Trotter Algorithm: Illustration*

.1.2.3.4 .1.2.4.3 .1.4.2.3 .4.1.2.3 4..1.3.2 .14..3.2 .1.34..2 .1.3.24.  
 .3.1.2.4 .3.1.4.2 .3.4.1.2 .4.3.1.2 4.3..2.1 3.4..2.1 3..24..1 3..2.14.  
 .23..1.4 .23..4.1 .2.43..1 .4.23..1 4..2.13. .24..13. .2.14.3. .2.13.4.

# Lexicographic Ordering of Permutations

- Recall that we assume the  $n$  objects of permutation are numbers

$$1, \dots, n.$$

- By treating a permutation as a base  $n + 1$  positional number, we can order the  $n!$  permutations  $12 \dots n, \dots, n \dots 21$  numerically, or lexicographically.

## The Immediate Successor of a Permutation

- In lexicographic ordering, the immediate successor of the permutation  $a_1 \cdots a_i \cdots a_n$  is  $b_1 \cdots b_i \cdots b_n$  if and only if the following holds.
- $a_1 = b_1, \dots, a_{i-1} = b_{i-1}$ .
- The subscript  $i$  is the largest  $i$  such that  $a_i < a_{i+1}$ . That is,

$$a_i < a_{i+1} > a_{i+2} > \cdots > a_n.$$

- $b_i = \min\{a_k : k > i, a_k > a_i\}$ .
- The sequence  $b_{i+1} \cdots b_n$  is an increasing sort of the numbers  $\{a_i, \dots, a_n\} \setminus \{b_i\}$ .

# *Lexicographically Ordered Permutations: Illustration*

1234 1243 1324 1342 1423 1432 2134 2143

2314 2341 2413 2431 3124 3142 3214 3241

3412 3421 4123 4132 4213 4231 4312 4321



## Generating Subsets

- There are at least three ways to generate the power set (the set of all subsets) of the set  $\{1, 2, \dots, n\}$ : decrease-by-1, bit-vector, and binary reflected Gray node.

## Generating Subsets: Decrease-By-One

- Let  $P(n)$  be the power set of  $\{1, \dots, n\}$ .
- The decrease-by-1 method is based on the observation that

$$P(n) = P(n - 1) \cup \{S \cup \{n\} : S \in P(n - 1)\}.$$

## Generating Subsets: Bit Vectors

- The members of the set  $S_n = \{1, \dots, n\}$  may be represented by the bits of a  $n$ -bit vector: bit position  $i$  represents member  $i + 1$ .
- Thus each of the  $2^n$  values  $0..2^n - 1$ , when written as  $n$ -bit base-2 numbers, gives a subset of  $S_n$ .
- Illustration:

0	1	2	3	4	5	6	7
000	001	010	011	100	101	110	111
$\{\}$	$\{1\}$	$\{2\}$	$\{1, 2\}$	$\{3\}$	$\{1, 3\}$	$\{2, 3\}$	$\{1, 2, 3\}$

## *Generating Subsets: Binary Reflected Gray Code*

- The binary reflected Gray code generates bit vectors that enjoys the minimal change property.

## *Decrease-by-a-Constant Factor Algorithms*

- Fake-coin identification problem
- Russian peasant method
- Josephus problem

## *Fake-Coin Identification Problem*

- Among  $n$  identically looking coins, one is fake and is known to be lighter.
- Use an uncalibrated scale, identify the fake coin.

## Fake-Coin Identification Problem

- One algorithm is to divide the  $n$  coins into at most three sets:  $\lfloor n/2 \rfloor$  coins,  $\lfloor n/2 \rfloor$  coins, and  $\lceil n/2 \rceil - \lfloor n/2 \rfloor$  coin.
- If  $n$  is even, weigh the only two sets and the lighter set contains the fake coin.
- If  $n$  is odd, weigh the first two sets. If they weigh the same, the third set contains the fake coin; else the lighter set contains the fake coin.
- The problem size is now reduced by about half and the same procedure can be applied again.

# An Analysis of the Fake-Coin Identification Algorithm

- The recurrence for the number of weighings  $W(n)$  is

$$W(n) = W(\lfloor n/2 \rfloor) + 1, \quad n > 1; \quad W(1) = 0$$

and the solution is  $W(n) = \lfloor \log_2 n \rfloor$ .

- It is known that a better algorithm with  $W(n) \approx \log_3 n$  if the coins are divided into three sets of size about  $n/3$ .



# The Russian Peasant Multiplication Method

- The product of two positive integers  $n$  and  $m$  can be found with halving, doubling, and adding:

$$nm = \begin{cases} \frac{n}{2} \times 2m, & n \text{ even;} \\ \frac{n-1}{2} \times 2m + m & n \text{ odd.} \end{cases}$$

- The operations of halving and doubling are simply 1-bit right-shift and 1-bit left-shift respectively.

# The Russian Peasant Multiplication Method: Illustration

$n$	$m$	
50	65	
25	130	130
12	260	
6	520	
3	1040	1040
1	2080	2080
		3250

# The Josephus Problem

- There are  $n$  persons numbered 1 to  $n$  standing in a circle.
- If  $n$  is even, eliminate all even-numbered persons;
- If  $n$  is odd, eliminate all even-numbered persons and the person numbered 1. The survivor who was numbered 3 becomes number 1.
- Repeat this process until a lone survivor is left.
- Who is the final survivor?

# The Solution to the Josephus Problem

- Let  $J(n)$  be the number of the final survivor.
- It can be verified that

$$J(2k) = 2J(k) - 1, \quad \text{and} \quad J(2k + 1) = 2J(k) + 1,$$

with the initial condition  $J(1) = 1$ .

- It turns out that

$$J(n) = \text{1-bit left cyclic shift of } n.$$

## *Variable-Size-Decrease Algorithms*

- Computing a median and the selection problem
- Interpolation search
- Searching and insertion in a binary search tree
- The game of NIM

# The Median and the Selection Problem

- The selection problem is about finding the  $k^{th}$  smallest element among  $n$  numbers.
- This number is called the  $k^{th}$  order statistic.
- When  $k = \lceil n/2 \rceil$ , the number is called the median of the given  $n$  numbers.
- For  $k = 1$  or  $k = n$ , we simply scan the list to find the minimum or maximum value respectively.
- To sort the entire list to find the  $k^{th}$  order statistics is doing too much work.

## Solving the Selection Problem by Partitioning

- An efficient algorithm (on average) is to use some partitioning algorithm like the one used in quicksort.
- Given an input array  $A[1..n]$  (note that index starts at 1), let  $A[s]$  be the pivot element.
- If  $s = k$ ,  $A[s]$  is the required answer.
- If  $s > k$ , then find the  $k^{th}$  smallest element among the  $s - 1$  elements of the left partition  $A[1..s - 1]$ .
- If  $s < k$ , then find the  $k - s^{th}$  smallest element among the  $n - s$  elements of the right partition  $A[s + 1..n]$ .

# An Analysis of the Selection by Partitioning Algorithm

- It is known that the average case efficiency is the same as if the size of array is reduced by about half in each iteration.
- If so the recurrence

$$C(n) = C(n/2) + \Theta(n)$$

and the solution is  $\Theta(n)$ .

- The worst case efficiency is the same as quicksort:  $\Theta(n^2)$ .



# Interpolation Search

- Recall that binary search looks for the key  $y$  among the partial array  $A[l..r]$  by comparing  $y$  with the array element indexed by  $\lfloor \frac{l+r}{2} \rfloor$ .
- Interpolation search assumes the keys are uniformly distributed so it looks for the key  $y$  among the partial array  $A[l..r]$  by comparing  $y$  with the array element indexed by

$$\lfloor x \rfloor = \left\lfloor l + \frac{(y - A[l])(r - l)}{A[r] - A[l]} \right\rfloor.$$

- The index  $\lfloor x \rfloor$  is obtained by assuming that the points  $(l, A[l])$ ,  $(x, y)$ ,  $(r, A[r])$  lie on a (straight) line.

# Interpolation Search

- If  $y \neq A[x]$ , the next iteration of the interpolation search narrows the search to  $A[l..x - 1]$  when  $y < A[x]$  or to  $A[x + 1..r]$  when  $A[x] < y$ .
- Empirically interpolation search performs better than binary search when the range of search is large.

## *Searching and Insertion in a Binary Search Tree*

- Recall that a binary search tree is either empty or the root key is larger than the keys in the left subtree but the smaller than the keys in the right subtree.
- The worst case time efficiency is  $\Theta(n)$  when the tree is badly balanced but the average case time efficiency is  $\Theta(\log n)$  as on the average the tree is reasonably balanced.

# Searching and Insertion in a Binary Search Tree

- The search is done recursively:

```
BS( key, T )           // looking for key in T
  if T = empty then return -1 fi
  if key = T.root then return T
  else
    if key < T.root then return BS(key, T.left)
    else
      return BS(key, T.right)
    fi
  fi
end
```

## *The Game of Nim: Rules*

- Consider the game Nim when two players take turn to remove chips from 2 or more piles of chips.
- Each time a player must remove at least one chip but at most all the chips of a pile.
- The player who removes the last pile of chips wins the game.

# The Game of Nim: A Fated Game

- It turns out that the winner is completely determined by the opening chip configurations and who moves first.
- To see this, assume there are  $k > 1$  piles of chips and the number of chips are  $n_1, \dots, n_k$  with the following binary representations:

$$\begin{array}{rcl} n_1 & = & b_{1,l} \quad \cdots \quad b_{1,0} \\ \vdots & = & \vdots \quad \ddots \quad \vdots \\ n_k & = & b_{k,l} \quad \cdots \quad b_{k,0} \\ \hline \oplus & = & b_l \quad \cdots \quad b_0 \\ \hline \hline \end{array}$$

where " $\oplus$ " denotes the bit-wise exclusively-or operation.

## The Game of Nim: A Fated Game

- If all the bits  $b_l, \dots, b_0$  are zero, the player who moves next will lose; otherwise, the player who moves next will win.
- When some of the bits  $b_l, \dots, b_0$  are not zero, from which pile and how many chips should the player remove?
- Let the left-most 1-bit be  $b_i$ .
- There must be at least a  $b_{j,i} = 1$  for some pile with  $n_j$  chips.
- For each of the bits  $b_l, \dots, b_0$  that is 1, flip the corresponding bits in  $n_j$  to obtain the number  $n'_j$ .
- Clearly,  $n'_j < n_j$ .
- The player should remove  $n_j - n'_j$  chips from pile  $j$ .

## *Take Away Message on Decrease-and-Conquer*

- The **decrease-and-conquer** technique exploits the relationship between a solution to a problem instance and a solution to a smaller instance of the same problem.
- The relationship may be exploited top-down with recursion or bottom-up with iteration. We may refer to the latter as **increase-and-conquer**.
- The three major variations are:
  - decrease by a constant (often 1),
  - decrease by a constant factor (often half), and
  - variable size decrease.