

## Computational Complexity

- Algorithmics vs. Complexity

- Algorithmics

- Given a problem, we can prove that the problem can be solved in a time of  $O(f(n))$  by giving and analyzing an algorithm
    - We'd like to reduce  $f(n)$  as much as possible

- Complexity

- It tells us any algorithm capable of solving our problem correctly takes a time  $\Omega(g(n))$
    - Now  $g(n)$  is a lower bound on the complexity of the problem
    - If  $f(n) \in \Theta(g(n))$ , we're satisfied

## Topics

- Arguing a lower bound

- Information-theoretic arguments
  - Adversary arguments

- Proof equivalence of complexity or compare complexity

- Linear reductions
  - NP-completeness

## Information-theoretic arguments

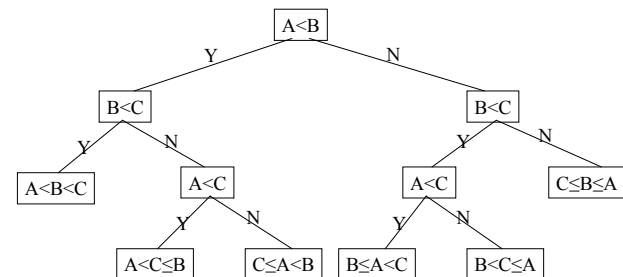
- Particularly applies to those problem involving comparisons

- Uses a decision tree to represent the working process of an algorithm on all possible data of a given size

- A decision tree is a binary tree where

- Each internal node contains a test on the data
    - Each leaf contains an output, called *verdict*
    - A *trip* through the tree starts from the root and recursively goes to the left subtree or the right subtree depending on whether the answer to the root is "yes" or "no"
    - The trip ends when it reaches a leaf (verdict)

## A decision tree for sorting



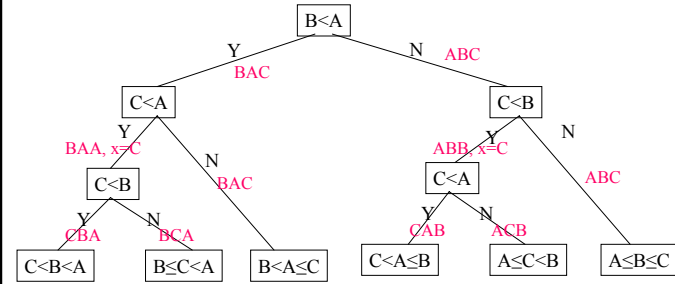
Note that the number of leaves = The number of possible outputs

## Insertion Sort

```
void insertionSort(int A[], int n)
{ // array index starts from 1 here
  int i, j, x;

  for (i=2; i<=n; i++) {
    x = A[i];
    j = i-1;
    while (j>0 && x<A[j]) {
      A[j+1] = A[j];
      j--;
    }
    A[j+1] = x;
  }
}
```

## A three-item insertion sort decision tree



## Observations

- The number possible outputs = The number of leaves (verdicts)
- The worst-case time is the height of the tree
- The average time is the average depth of leaves assuming equal distribution

## Theorem

- Any binary tree with  $k$  leaves has an average height of at least  $\lg k$ 
  - Let  $h(k)$  be smallest possible total depths for a tree of  $k$  leaves

$$h(k) = \begin{cases} 0 & k \leq 1 \\ \min_{1 \leq i \leq k-1} (h(i) + h(k-i) + k) & \text{otherwise} \end{cases}$$

- Any comparison-based algorithm takes a worst case time and average case time  $\Omega(n \log n)$

### Adversary arguments

- Start the algorithm on an input that is initially unspecified except for its size
- When the algorithm probes the input, the malevolent daemon, the adversary, answers in a way that will force the algorithm to work hard
  - The daemon's goal is to keep the algorithm uncertain of the correct answer as long as possible
  - Constraint: the daemon's answers must be consistent—there always exists at least one input that could cause the algorithm to see exactly the same answers on its probes

### Finding the maximum of an array

- A simple algorithm takes a time of  $\Theta(n)$
- Using decision tree, we get
  - Any comparison based algorithm to find the maximum must perform at least  $\lceil \lg n \rceil$  comparisons in worst case
- Can we find a tighter lower bound?
  - We use adversary arguments to show  $n-1$  comparisons are necessary

### Find the median

- For a comparison-based algorithm, we can easily argue a lower bound of  $\left\lceil \frac{n}{2} \right\rceil$  comparisons
- Can we find a tighter bound?
  - We can prove a lower bound of  $\frac{3(n-1)}{2}$  comparisons

### Adversary arguments

- Assume  $n$  is odd and  $n \geq 3$ .
- Initially, the daemon sets each entry to “uninitialized”
  - As the algorithm makes comparisons, the daemon set values between 1 and  $n$  (low) or between  $3n+1$  to  $4n$  (high).
  - The daemon makes sure #low items = #high items

## Comparison

- When  $T[i]$  is asked to compare to  $T[j]$ 
  1.  $T[i]$  and  $T[j]$  are both uninitialized
    - $T[i]$  is set to  $i$  and  $T[j]$  is set to  $3n+j$ .
  2. One is uninitialized
    1. If it is the only one uninitialized item left, set its value to  $2n$  which becomes *provisional median*.
    2. If  $T[i]$  is low, set  $T[j]$  to high value  $3n+j$ . And add one uninitialized item  $T[k]$  to low, i.e., set  $T[k]$  to  $k$ .
    3. If  $T[j]$  is low, set  $T[i]$  to high value  $3n+i$ . And add one uninitialized item  $T[k]$  to low, i.e., set  $T[k]$  to  $k$ .
    4. If  $T[i]$  is high, set  $T[j]$  to low value  $j$ . And add one uninitialized item  $T[k]$  to high, i.e., set  $T[k]$  to  $3n+k$ .
    5. If  $T[j]$  is high, set  $T[i]$  to low value  $i$ . And add one uninitialized item  $T[k]$  to high, i.e., set  $T[k]$  to  $3n+k$ .
  3. Both are initialized

## Arguments

- Both are initialized
  - If both are low or one low and one provisional median, then the smaller has *lost* a comparison
  - If both are high or one high and one provisional median, then the larger has *lost* a comparison
  - Otherwise, no one lost comparison
- Arguments: if less than  $3(n-1)/2$  comparisons
  - $(n-1)/2$  comparisons needed to initialize all values
  - Less than  $n-1$  comparisons for initialized values
    - Less than  $n-1$  values lost comparisons
    - At least one item in addition to the provisional median never lost a comparison