

# IE 495 Lecture 12

October 5, 2000

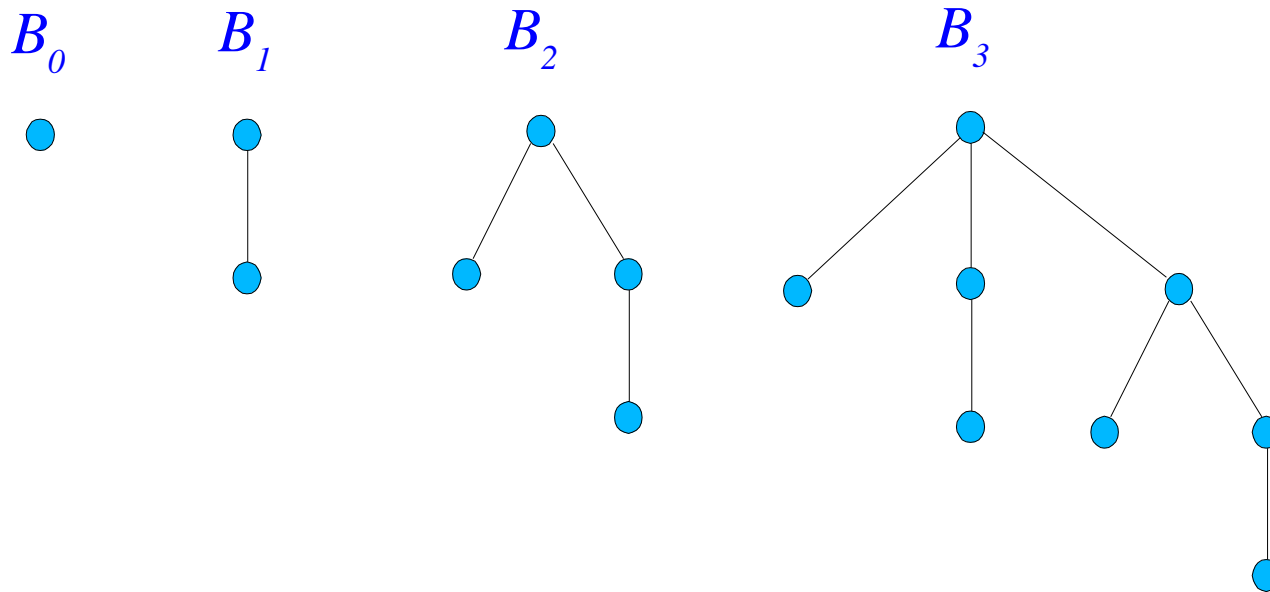
# Reading for This Lecture

- Primary
  - Horowitz and Sahni, Chapter 2, Section 3
  - Kozen, Lectures 8-11

# Review From Last Time

# Binomial Trees

- The *binomial tree* of rank  $i$  ( $B_i$ ) is defined recursively.
- $B_i$  consists of a *root* with  $i$  children  $B_0, \dots, B_{i-1}$ .



# Binomial Heaps

- A *binomial heap* is a collection of heap ordered binomial trees and a pointer to the overall max/min.
- No more than one tree of each rank is allowed.
- The children of each vertex are maintained in a circular linked list.
- The basic operation is *linking*.
- Two trees of rank  $i$  can be combined into one tree of rank  $i+1$  in constant time.

# Eager Meld

- We can combine two heaps by performing a *meld()* reminiscent of binary addition.
- Successively link trees of equal rank and "carry" one if necessary.
- Must track the position of the new min/max element.
- This operation takes  $O(\log n)$  time.

# Inserting into a Binomial Heap

- To *insert()* an element:
  - Make a new heap from the single element to be inserted.
  - Meld the new heap with the old one.
- To *make\_heap()* from scratch, perform a sequence of inserts.
- To *delete()* the min/max element:
  - The children of this element form a new binomial heap.
  - Meld the old heap and the new one.

# Amortized Analysis

- *meld()* and *delete()* both take  $O(\log n)$ .
- We will use *amortized analysis* to show that *insert()* is constant time overall.
- Idea: The total number of linking operations can never be more than the number of insert operations.
- This means that any sequence of inserts takes constant time *on average*.

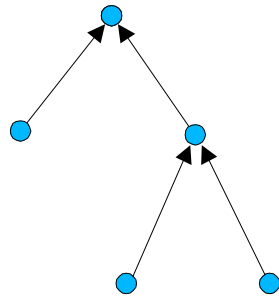


# Data Structures for Disjoint Sets

- We have a set  $S$  and a partition  $S_1, \dots, S_n$  of  $S$ .
- We want a data structure that supports
  - `union()`
  - `find()`
- Applications
  - Constructing equivalence classes
  - Graph algorithms

# Union-Find

- Represent each member of the partition as a rooted tree.
- Choose a designated "representative".
- All other elements are connected to the representative.



# First implementation

- `union()`
  - Point root of set  $A$  to root of set  $B$
- `find()`
  - Follow the path to the root.
- Analysis

# A Tale of Two Heuristics

- How can we improve the complexity of *find()*?
- Heuristic 1
- Heuristic 2

# Analysis

- Heuristic 1 guarantees that the depth of each tree is no more than  $\lfloor \log n \rfloor + 1$ .
- The proof of this is by induction.
- This implies that *find()* can be performed in  $O(\log n)$
- Heuristic 2 allows us to perform *find()* in *almost* constant time (amortized).

# Ackerman's Function

- Ackerman's function is an extremely fast growing function.
- Definition
  - $A_0(x) = x + 1$
  - $A_{k+1}(x) = A_k^x(x)$ , where  $A_k^{i+1}(x) = A_k(A_k^i(x))$
- $A_0(x) = x+1$ ,  $A_1(x) = 2^x$ ,  $A_2(x) = x2^x$ ,  $A_3(x) \geq 2 \uparrow x$
- $A_4(2)$  is greater than the number of particles in the known universe or the number of nanoseconds since the Big Bang (large number).

# Inverse Ackerman's Function

- Define  $A(k) = A_k(2)$ .
- Now define  $\alpha(n) =$  smallest  $k$  such that  $A(k) \geq n$
- $\alpha(n)$  is the inverse Ackerman's function
- $\alpha(n)$  is 4 for all practical purposes.
- Let  $T(m, n)$  be the running time of a sequence of  $m \geq n$  *find()* operations and  $n-1$  *union()* operations.
- $T(m, n) \in O(\alpha(n)(m+n))$

# Hash Tables

- Symbol Table
  - Determine presence of an arbitrary element
  - Allow easy insertion and deletion
- Hashing is an easy and efficient implementation
- Hash function
  - Maps each possible element into a specified bucket
  - The number of buckets is much less than the number of possible elements
  - Each bucket can store a limited number of elements



# Parameters

- $T$  = total number of possible elements
- $b$  = number of buckets
- $s$  = number of elements allowed in each bucket
- $n$  = number of elements in the table
- $n/T$  = element density
- $\alpha = n/sb$  = loading density

# Hash Functions

- *Collision*: two elements map to the same bucket
- *Overflow*: too many elements in one bucket
- Choosing a hash function
  - easy to compute
  - minimize collisions
- If  $P(f(X) = i) = 1/b$  over all elements  $X$ , then  $f$  is a *uniform hash function*

# Sample Hash Function

- Interpret the element of the set as an integer  $X$
- Take the hash function to

$$f(X) = X \bmod M$$

- $M$  is the number of buckets
- The choice of  $M$  is critical
- $M$  should not be a power of 2 or an even number
- $M$  should be a prime number with some other nice properties

# Overflow Handling

- Use the next available slot
  - Bad performance when the hash table fills up.
  - Can end up searching the whole table.
  - Average number of comparisons  $(2-\alpha)/(2-2\alpha)$ .
- Use linked lists
  - Only compare items with same hash value.
  - Average number of comparison  $1 + \alpha/2$ .
- Average case for hash tables is good, but worst case is very bad.