

MA/CSSE 473

Day 29

Day30-Dynamic Binomial-Warshall

Dynamic
Programming

Binomial Coefficients

Warshall's algorithm



B-trees (Section 2 only)

- We will do a quick overview here.
- For the whole scoop on B-trees (Actually B+ trees), take CSSE 433, Advanced Databases.
- Nodes can contain multiple keys and pointers to other subtrees



B-tree nodes

- Each node can represent a block of disk storage; pointers are disk addresses
- This way, when we look up a node (requiring a disk access), we can get a lot more information than if we used a binary tree
- In an n -node of a B-tree, there are n pointers to subtrees, and thus $n-1$ keys
- All keys in T_i are $\geq K_i$ and $< K_{i+1}$
 K_i is the smallest key that appears in T_i

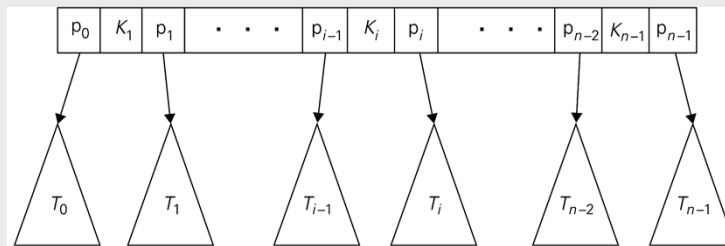


FIGURE 7.7 Parental node of a B-tree

B-tree nodes (tree of order m)

- All nodes have at most $m-1$ keys
- All keys and associated data are stored in special **leaf** nodes (that thus need no child pointers)
- The other (parent) nodes are **index** nodes
- All index nodes except the root have between $\lceil m/2 \rceil$ and m children
- root has between 2 and m children
- All leaves are at the same level
- The space-time tradeoff is because of duplicating some keys at multiple levels of the tree
- Especially useful for **data that is too big to fit in memory**. Why?
- Example on next slide



Example B-tree(order 4)

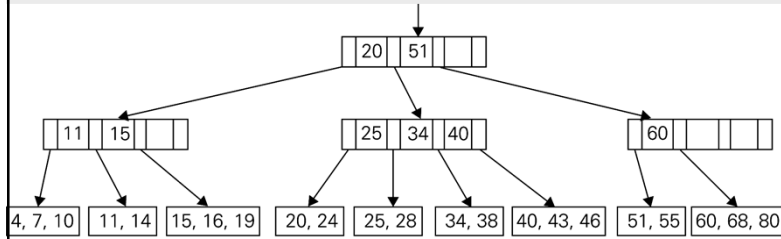


FIGURE 7.8 Example of a B-tree of order 4

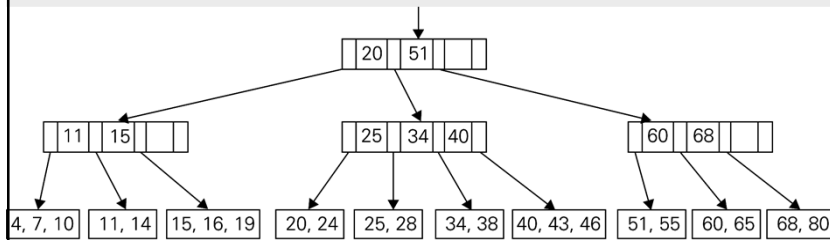


FIGURE 7.9 B-tree obtained after inserting 65 into the B-tree in Figure 7.8

B-tree Animation

- <http://slady.net/java/bt/view.php?w=800&h=600>

Search for an item

- Within each parent or leaf node, the items are sorted, so we can use binary search ($\log m$), which is a constant with respect to n , the number of items in the table
- Thus the search time is proportional to the height of the tree
- Max height is approximately $\log_{\lceil m/2 \rceil} n$
- **Exercise for you:** Read and understand the straightforward analysis on pages 273-274
- Insert and delete are also proportional to height of the tree



Preview: Dynamic programming

- Used for problems with overlapping subproblems
- Typically, we save (memoize) solutions to the subproblems, to avoid recomputing them.



Dynamic Programming Example

- Binomial Coefficients:
- $C(n,k)$ is the coefficient of x^k in the expansion of $(1+x)^n$
- $C(n,0) = C(n,n) = 1$.
- If $0 < k < n$, $C(n, k) = C(n-1, k) + C(n-1, k-1)$
- Can show by induction that the "usual" factorial formula for $C(n, k)$ follows from this definition.
 - Let's do it together
- If we don't cache values as we compute them, this can take a lot of time, because of duplicate (overlapping) computation.



Computing a binomial coefficient

Binomial coefficients are coefficients of the binomial formula:

$$(a + b)^n = C(n,0)a^n b^0 + \dots + C(n,k)a^{n-k}b^k + \dots + C(n,n)a^0 b^n$$

Recurrence: $C(n,k) = C(n-1,k) + C(n-1,k-1)$ for $n > k > 0$

$$C(n,0) = 1, \quad C(n,n) = 1 \quad \text{for } n \geq 0$$

Value of $C(n,k)$ can be computed by filling a table:

	0	1	2	...	k-1	k
0	1					
1	1	1				
.						
.						
.						
n-1					$C(n-1,k-1)$	$C(n-1,k)$
n						$C(n,k)$



Computing $C(n, k)$:

```

ALGORITHM Binomial( $n, k$ )
    //Computes  $C(n, k)$  by the dynamic programming algorithm
    //Input: A pair of nonnegative integers  $n \geq k \geq 0$ 
    //Output: The value of  $C(n, k)$ 
    for  $i \leftarrow 0$  to  $n$  do
        for  $j \leftarrow 0$  to  $\min(i, k)$  do
            if  $j = 0$  or  $j = i$ 
                 $C[i, j] \leftarrow 1$ 
            else  $C[i, j] \leftarrow C[i - 1, j - 1] + C[i - 1, j]$ 
    return  $C[n, k]$ 
    
```

Time efficiency: $\Theta(nk)$

Space efficiency: $\Theta(nk)$

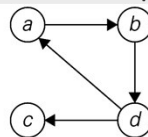
Exercise 8.1.7 asks you to compare the efficiency of this approach with some other approaches

If we are computing $C(n, k)$ for many different n and k values, we could cache the table between calls.



Transitive closure of a directed graph

- For each pair of vertices, (A, B) , in the directed graph G , is there a path from A to B in G ?
- Start with the boolean adjacency matrix A for the n -node graph G . $A[i][j]$ is 1 if and only if G has a directed edge from node i to node j .
- The **transitive closure** of G is the boolean matrix T such that $T[i][j]$ is 1 iff there is a nontrivial directed path from node i to node j in G .
- If we use boolean adjacency matrices, what does M^2 represent? M^3 ?
- In boolean matrix multiplication, $+$ stands for **or**, and $*$ stands for **and**



(a)

$$A = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

(b)

$$T = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

(c)

FIGURE 8.2 (a) Digraph. (b) Its adjacency matrix. (c) Its transitive closure.

Transitive closure *via* multiplication

- Again, using + for **or**, we get
$$T = M + M^2 + M^3 + \dots$$
- Can we limit it to a finite operation?
- We can stop at M^{n-1} .
 - How do we know this?
- Number of numeric multiplications for solving the whole problem?



Warshall's algorithm

- Similar to binomial coefficients algorithm
- Assumes that the vertices have been numbered 1, 2, ..., n
- Define the boolean matrix $R^{(k)}$ as follows:
 - $R^{(k)}[i][j]$ is 1 iff there is a path in the directed graph $i=v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_s=j$, where
 - $s \geq 1$, and
 - for all $t = 1, \dots, s-1, v_t \leq k$
i.e, none of the intermediate vertices are numbered higher than k .
- Note that T is $R^{(n)}$



R^(k) example

- R^(k)_{[i][j]} is 1 iff there is a path in the directed graph

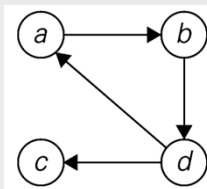
$i=v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_s=j$, where

– $s > 1$, and

– for all $t = 2, \dots, s-1, v_t \leq k$

- assuming that the nodes are numbered in alphabetical order, calculate R⁽⁰⁾ and R⁽¹⁾

You can find a larger example in a book that available at Safari Books on-line, through the Logan Library Web page (in the Databases section near the top of the page). The book is Sedgwick, *Algorithms Part 5*. See section 19-3



$$A = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$



Quickly Calculating R^(k)

- Back to the matrix multiplication approach:
 - How much time did it take to compute $A^k[i][j]$, once we have A^{k-1} ?
- Can we do better when calculating R^(k)_{[i][j]} from R^(k-1)?
- How can R^(k)_{[i][j]} be 1?
 - either R^(k-1)_{[i][j]} is 1, or
 - there is a path from i to k that uses no vertices higher than k-1, and a similar path from k to j.
- Thus R^(k)_{[i][j]} = R^(k-1)_{[i][j]} **or** (R^(k-1)_{[i][k]} **and** R^(k-1)_{[k][j]})
- Note that this can be calculated in constant time
- Time for calculating R^(k) from R^(k-1)?
- Total time for Warshall's algorithm?
- How does this time compare to using DFS?

Code and example on next slides



ALGORITHM *Warshall*($A[1..n, 1..n]$)
 //Implements Warshall's algorithm for computing the transitive closure
 //Input: The adjacency matrix A of a digraph with n vertices
 //Output: The transitive closure of the digraph
 $R^{(0)} \leftarrow A$
for $k \leftarrow 1$ **to** n **do**
 for $i \leftarrow 1$ **to** n **do**
 for $j \leftarrow 1$ **to** n **do**
 $R^{(k)}[i, j] \leftarrow R^{(k-1)}[i, j]$ **or** ($R^{(k-1)}[i, k]$ **and** $R^{(k-1)}[k, j]$)
return $R^{(n)}$

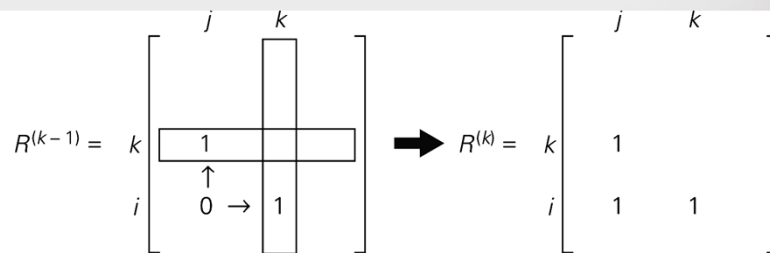


FIGURE 8.3 Rule for changing zeros in Warshall's algorithm

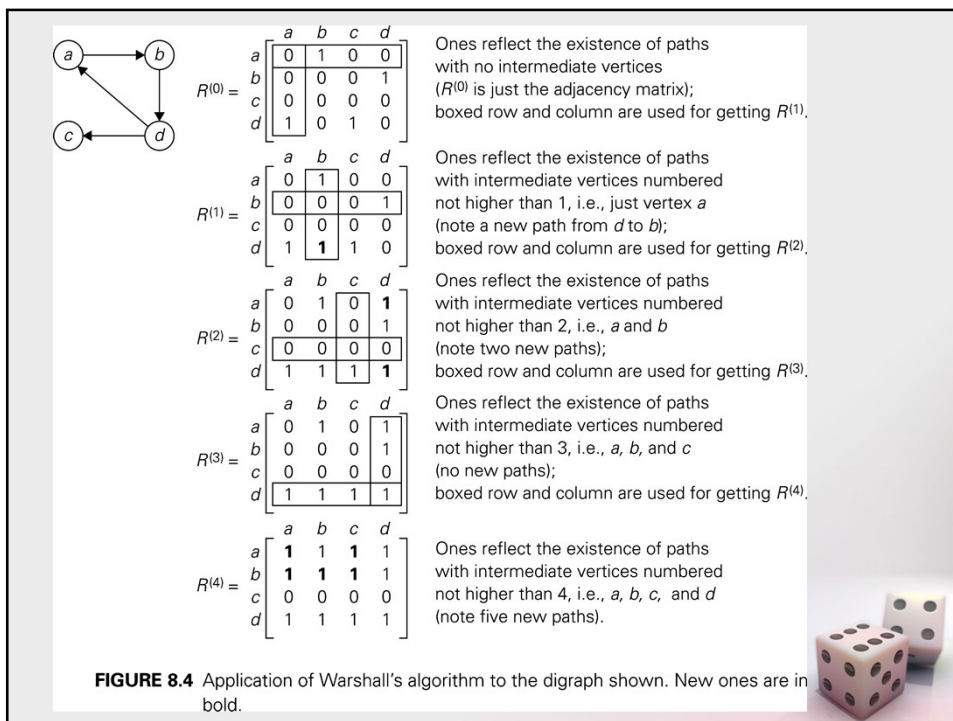


FIGURE 8.4 Application of Warshall's algorithm to the digraph shown. New ones are in bold.