The Longest-Common-Subsequence Problem (1/3)

- Given a sequence $X = \langle x_1, x_2, \dots, x_m \rangle$, another sequence $Z = \langle z_1, z_2, \dots, z_k \rangle$ is a subsequence of X if there exists a strictly increasing sequence $\langle i_1, i_2, \dots, i_k \rangle$ of indices of X such that for all j = 1, $2, \dots, k$, we have $x_{i_j} = z_j$.
 - For example, $Z = \langle B, C, D, B \rangle$ is a subsequence of $X = \langle A, B, C, B, D, A, B \rangle$ with corresponding index sequence $\langle 2, 3, 5, 7 \rangle$.
- Given two sequences X and Y, we say that a sequence Z is a common subsequence of X and Y if Z is a subsequence of both X and Y.

The Longest-Common-Subsequence Problem (2/3)

- For example, if $X = \langle A, B, C, B, D, A, B \rangle$ and $Y = \langle B, D, C, A, B, A \rangle$, the sequence $\langle B, C, A \rangle$ is a common subsequence of both X and Y.
- The sequence <B, C, A> is not a longest common subsequence (LCS) of X and Y, however, since it has length 3 and the sequence <B, C, B, A>, which is also common to both X and Y, has length 4.
- The sequence <B, C, B, A> is an LCS of X and Y, as is the sequence <B, D, A, B>, since X and Y have no common subsequence of length 5 or greater.

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The Longest-Common-Subsequence Problem (3/3)

- In the longest-common-subsequence problem, we are given two sequences $X = \langle x_1, x_2, \dots, x_m \rangle$ and $Y = \langle y_1, y_2, \dots, y_n \rangle$ and wish to find a maximum-length common subsequence of X and Y.
- Example1: $S_1 = AGCGTAG$, $S_2 = GTCAGA$, please find a maximum-length common subsequence.

Step 1: Characterizing a Longest Common Subsequence (1/2)

- In a brute-force approach to solving the LCS problem, we would
 - enumerate all subsequences of X (2^m subsequences in total) and
 - check each subsequence to see whether it is also a subsequence of Y,
 - keeping track of the longest subsequence we find.
- Requires exponential time, impractical for long sequences.

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Step 1: Characterizing a Longest Common Subsequence (2/2)

- Given a sequence $X = \langle x_1, x_2, ..., x_m \rangle$, we define the i th prefix of X, for i = 0, 1, ..., m, as $X_i = \langle x_1, x_2, ..., x_i \rangle$.
 - For example, if $X = \langle A, B, C, B, D, A, B \rangle$, then $X_4 = \langle A, B, C, B \rangle$ and X_0 is the empty sequence.

Theorem 15.1 (Optimal substructure of an LCS)

Let $X = \langle x_1, x_2, \dots, x_m \rangle$ and $Y = \langle y_1, y_2, \dots, y_n \rangle$ be sequences, and let $Z = \langle z_1, z_2, \dots, z_k \rangle$ be any LCS of X and Y.

- 1. If $x_m = y_n$, then $z_k = x_m = y_n$ and Z_{k-1} is an LCS of X_{m-1} and Y_{n-1} .
- 2. If $x_m \neq y_n$, then $z_k \neq x_m$ implies that Z is an LCS of X_{m-1} and Y.
- 3. If $x_m \neq y_n$, then $z_k \neq y_n$ implies that Z is an LCS of X and Y_{n-1} .

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Step 2: A Recursive Solution (1/2)

- **Theorem 15.1** implies that we should examine either one or two subproblems when finding an LCS of $X = \langle x_1, x_2, \dots, x_m \rangle$ and $Y = \langle y_1, y_2, \dots, y_n \rangle$.
 - If $x_m = y_n$, we must find an LCS of X_{m-1} and Y_{n-1} . Appending $x_m = y_n$ to this LCS yields an LCS of X and Y.
 - If $x_m \neq y_n$, then we must solve two subproblems: finding an LCS of X_{m-1} and Y and finding an LCS of X and Y_{n-1} . Whichever of these two LCSs is longer is an LCS of X and Y.

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Step 2: A Recursive Solution (2/2)

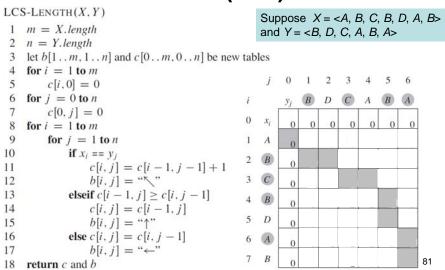
 Let us define c[i, j] to be the length of an LCS of the sequences X_i and Y_j.

$$c[i,j] = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0, \\ c[i-1,j-1]+1 & \text{if } i,j > 0 \text{ and } x_i = y_j, \\ \max(c[i,j-1],c[i-1,j]) & \text{if } i,j > 0 \text{ and } x_i \neq y_j. \end{cases}$$
(15.9)

Step 3: Computing the Length of an LCS (1/10)

- Procedure LCS-LENGTH takes two sequences $X = \langle x_1, x_2, ..., x_m \rangle$ and $Y = \langle y_1, y_2, ..., y_n \rangle$ as inputs.
 - It stores the c[i, j] values in a table c[0..m, 0..n], and it computes the entries in row-major order.
 - The procedure also maintains the table b[1..m, 1..n] to help us construct an optimal solution.
 - The procedure returns the b and c tables; c[m, n] contains the length of an LCS of X and Y.

Step 3: Computing the Length of an LCS (2/10)

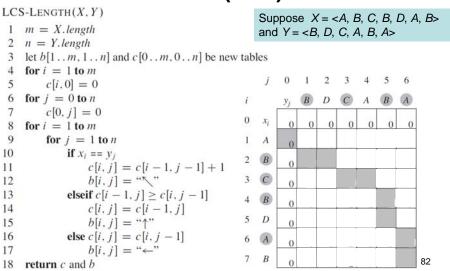


11 12 13 14

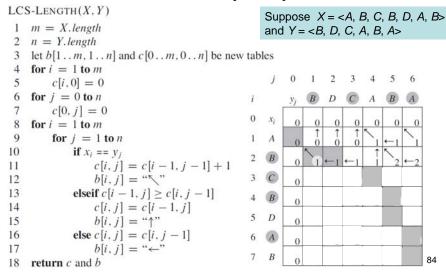
Step 3: Computing the Length of an LCS (4/10)

```
LCS-LENGTH(X, Y)
                                                    Suppose X = \langle A, B, C, B, D, A, B \rangle
 1 m = X.length
                                                    and Y = \langle B, D, C, A, B, A \rangle
 2 \quad n = Y.length
 3 let b[1..m, 1..n] and c[0..m, 0..n] be new tables
    for i = 1 to m
                                                            1 2 3 4 5 6
         c[i, 0] = 0
    for j = 0 to n
                                                                 D C A B A
         c[0, j] = 0
    for i = 1 to m
         for i = 1 to n
             if x_i == y_i
                                                 2 B
11
                 c[i, j] = c[i-1, j-1] + 1
                                                 3 C
12
                 b[i,j] = "\"
13
             elseif c[i - 1, j] \ge c[i, j - 1]
14
                 c[i, j] = c[i - 1, j]
                                                 5 D
15
                 b[i, j] = "\uparrow"
16
             else c[i, j] = c[i, j - 1]
                                                    A
                                                 6
17
                  b[i, j] = "\leftarrow"
                                                     B
                                                                                      83
18 return c and b
```

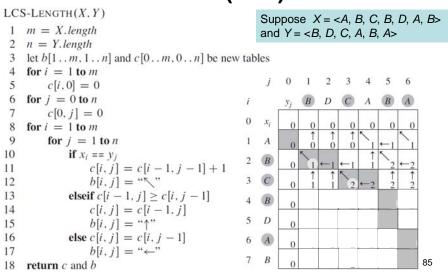
Step 3: Computing the Length of an LCS (3/10)



Step 3: Computing the Length of an LCS (5/10)



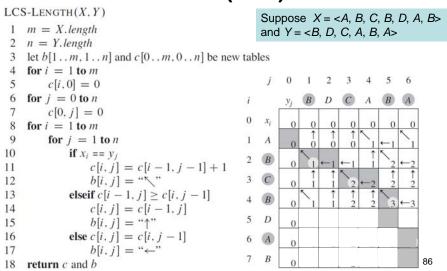
Step 3: Computing the Length of an LCS (6/10)



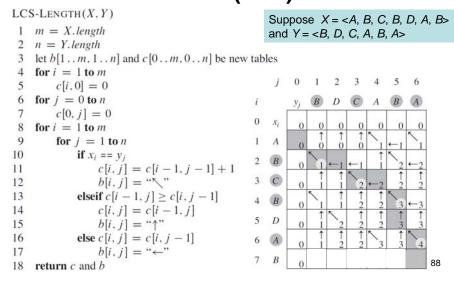
Step 3: Computing the Length of an LCS (8/10)

```
LCS-LENGTH(X, Y)
                                                     Suppose X = \langle A, B, C, B, D, A, B \rangle
 1 m = X.length
                                                     and Y = \langle B, D, C, A, B, A \rangle
 2 \quad n = Y.length
 3 let b[1..m, 1..n] and c[0..m, 0..n] be new tables
    for i = 1 to m
                                                             1 2 3 4 5 6
         c[i, 0] = 0
    for j = 0 to n
         c[0, j] = 0
    for i = 1 to m
         for i = 1 to n
             if x_i == y_i
                                                  2 B
11
                  c[i, j] = c[i-1, j-1] + 1
                                                  3 C
12
                  b[i,j] = "\"
13
             elseif c[i - 1, j] \ge c[i, j - 1]
14
                  c[i, j] = c[i - 1, j]
                                                  5 D
15
                  b[i, j] = "\uparrow"
16
             else c[i, j] = c[i, j - 1]
                                                     A
                                                  6
17
                  b[i, j] = "\leftarrow"
                                                  7
                                                      B
                                                                                       87
18 return c and b
```

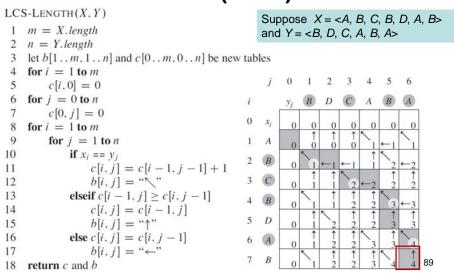
Step 3: Computing the Length of an LCS (7/10)



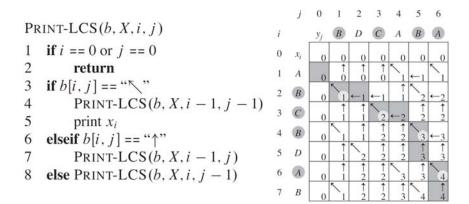
Step 3: Computing the Length of an LCS (9/10)



Step 3: Computing the Length of an LCS (10/10)



Step 4: Constructing an LCS



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Outine

- Rod Cutting
- Matrix-Chain Multiplication
- Elements of Dynamic Programming
- Longest Common Subsequence
- Optimal Binary Search Trees

Introduction (1/2)

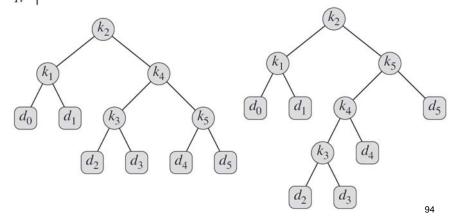
- Suppose that we are designing a program to translate text from English to Chinese.
 - We could perform lookup operations by building a binary search tree with n English words as keys.
 - Since the number of nodes visited when searching for a key in a binary search tree equals one plus the depth of the node containing the key, we want words that occur frequently in the text to be placed nearer the root.

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Introduction (2/2)

- We are given a sequence $K = \langle k_1, k_2, ..., k_n \rangle$ of n distinct keys in sorted order $(k_1 < k_2 < ... < k_n)$, and we wish to build a binary search tree from these keys.
- For each key k_i , we have a probability p_i that a search will be for k_i .
- Some searches may be for values not in K, and so we also have n+1 "dummy keys" d_0 , d_1 , d_2 , ..., d_n representing values not in K.
- For each dummy key d_i , we have a probability q_i that a search will correspond to d_i .

Two Binary Search Trees



Expected Cost of a Search in T (1/2)

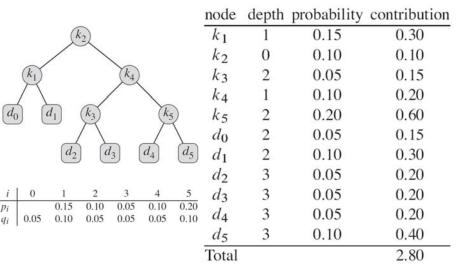
Let us assume that the actual cost of a search equals the number of nodes examined, i.e., the depth of the node found by the search in *T*, plus
1. Then the expected cost of a search in *T* is

$$E[\operatorname{search cost in} T] = \sum_{i=1}^{n} (\operatorname{depth}_{T}(k_{i}) + 1) \cdot p_{i} + \sum_{i=0}^{n} (\operatorname{depth}_{T}(d_{i}) + 1) \cdot q_{i}$$

$$= 1 + \sum_{i=1}^{n} \operatorname{depth}_{T}(k_{i}) \cdot p_{i} + \sum_{i=0}^{n} \operatorname{depth}_{T}(d_{i}) \cdot q_{i} ,$$

where $depth_T$ denotes a node's depth in the tree T.

Expected Cost of a Search in T (2/2)



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