

CS146: Data Structures and Algorithms

Lecture 19



DYNAMIC PROGRAMMING
ALL PAIRS SHORTEST SOURCE SHORTEST PATH
(CH 25)
LONGEST COMMON SUBSEQUENCE (CH 15)

INSTRUCTOR: KATERINA POTIKA
CS SJSU

DP: Three-step method

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1. Define subproblems
2. Write down the recurrence that relates subproblems
3. Recognize and solve the base cases

All pairs shortest path (Ch 25)

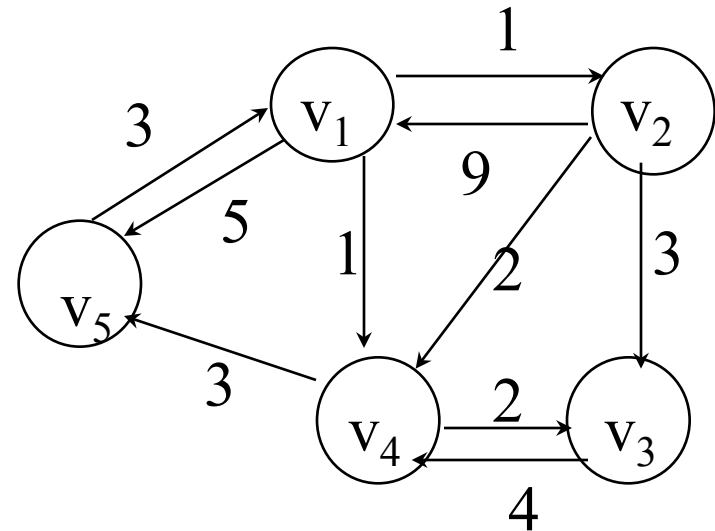
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- Input: A weighted graph (may contain negative edges but no negative cycles)
- Output: the shortest path between every pair of vertices of the graph
- *A representation*: a weight matrix where
 - $W(i,j)=0$ if $i=j$.
 - $W(i,j)=\infty$ if there is no edge between i and j .
 - $W(i,j)$ =“weight of edge”
- Note: we have shown principle of optimality applies to shortest path problems

The weight matrix and the graph

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	1	2	3	4	5
1	0	1	∞	1	5
2	9	0	3	2	∞
3	∞	∞	0	4	∞
4	∞	∞	2	0	3
5	3	∞	∞	∞	0



The subproblems

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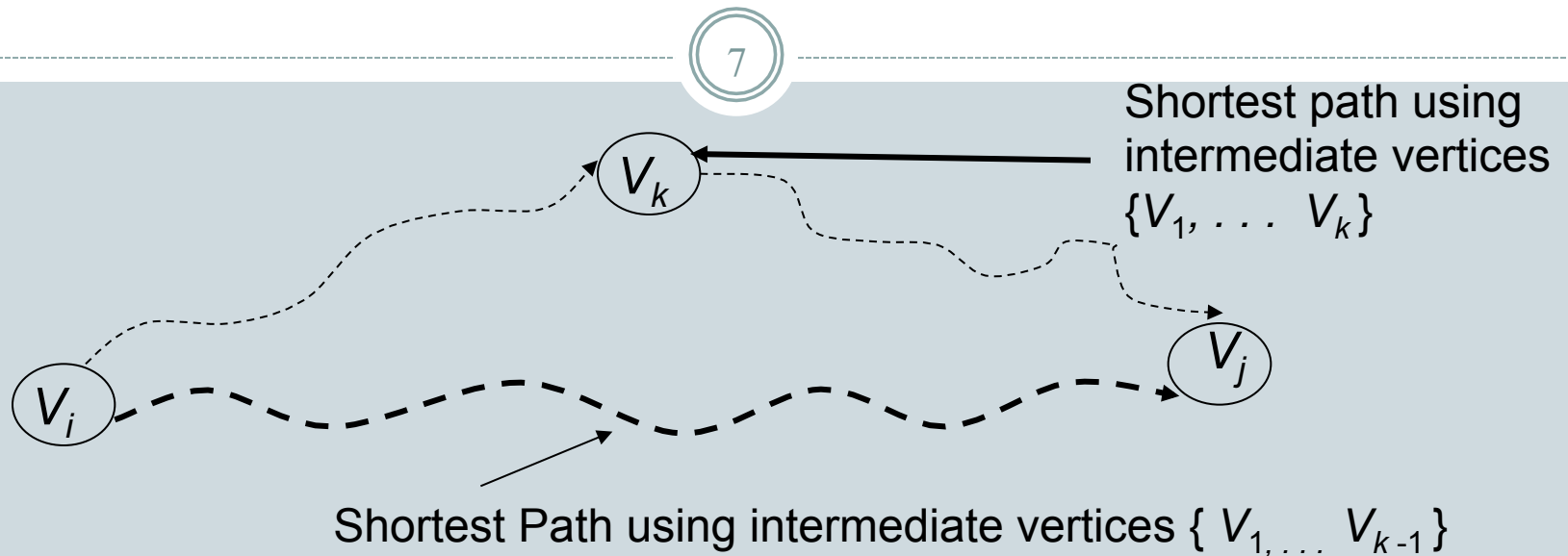
- How can we define the shortest distance $d_{i,j}$ in terms of “smaller” problems?
- One way is to restrict the paths to only include vertices from a *restricted* subset.
- Initially, the subset is empty.
- Then, it is incrementally increased until it includes all the vertices.

The subproblems

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- Let $D^{(k)}[i,j]$ = weight of a shortest path from v_i to v_j using only vertices from $\{v_1, v_2, \dots, v_k\}$ as intermediate vertices in the path
 - $D^{(0)} = W$
 - $D^{(n)} = D$ which is the goal matrix
- How do we compute $D^{(k)}$ from $D^{(k-1)}$?

The Recursive Definition



A shortest path from v_i to v_j restricted to using only vertices from $\{v_1, v_2, \dots, v_k\}$ as intermediate vertices

- [Case1] does **not use** v_k . Then $D^{(k)}[i,j] = D^{(k-1)}[i,j]$
- [Case2] does **use** v_k . Then $D^{(k)}[i,j] = D^{(k-1)}[i,k] + D^{(k-1)}[k,j]$.

The recursive definition

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- Since

$$D^{(k)}[i,j] = D^{(k-1)}[i,j] \text{ or}$$

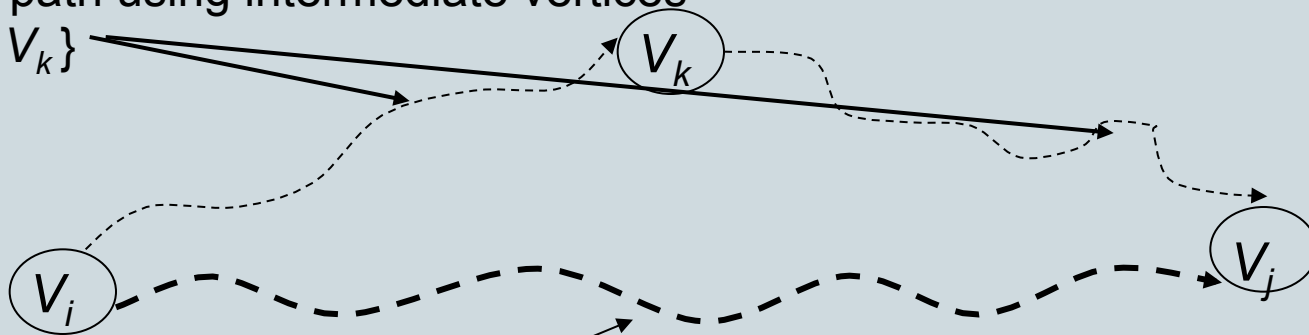
$$D^{(k)}[i,j] = D^{(k-1)}[i,k] + D^{(k-1)}[k,j].$$

We conclude:

$$D^{(k)}[i,j] = \min\{ D^{(k-1)}[i,j], D^{(k-1)}[i,k] + D^{(k-1)}[k,j] \}.$$

Shortest path using intermediate vertices

$\{V_1, \dots, V_k\}$



Shortest Path using intermediate vertices $\{V_1, \dots, V_{k-1}\}$

The pointer array Π

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- Used to enable finding a shortest path (solution)
- Initially the array contains 0
- Each time that a shorter path from i to j is found the k that provided the minimum is saved (highest index node on the path from i to j)
- To print the intermediate nodes on the shortest path a recursive procedure that print the shortest paths from i and k , and from k to j can be used

Floyd's Algorithm Using $(n+1)$ D matrices

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FLOYD-WARSHALL(W, n)

$D^{(0)} \leftarrow W$

for $k \leftarrow 1$ **to** n

do for $i \leftarrow 1$ **to** n

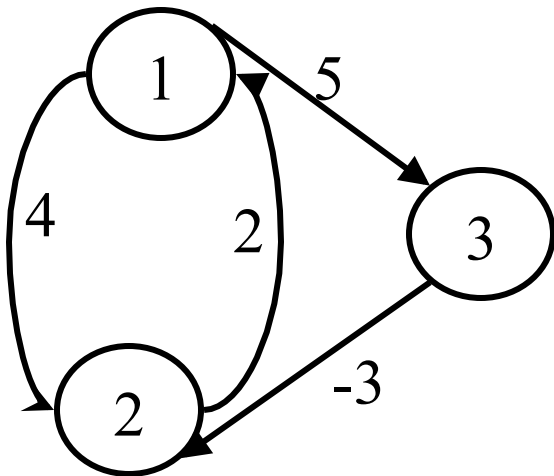
do for $j \leftarrow 1$ **to** n

do $d_{ij}^{(k)} \leftarrow \min(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)})$

return $D^{(n)}$

Example

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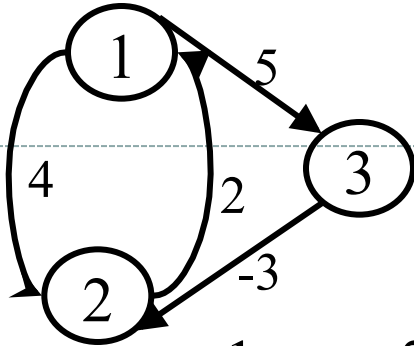
$$W = D^0 =$$

	1	2	3
1	0	4	5
2	2	0	∞
3	∞	-3	0

$$\Pi =$$

	1	2	3
1	0	0	0
2	0	0	0
3	0	0	0

k = 1 Vertex 1 can be intermediate node



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$$D^0 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{array}{|c|c|c|} \hline 0 & 4 & 5 \\ \hline 2 & 0 & \infty \\ \hline \infty & -3 & 0 \\ \hline \end{array} \end{matrix}$$

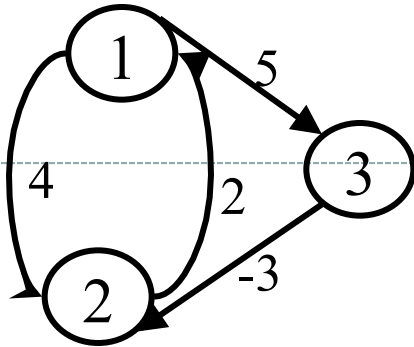
$$D^1 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{array}{|c|c|c|} \hline 0 & 4 & 5 \\ \hline 2 & 0 & 7 \\ \hline \infty & -3 & 0 \\ \hline \end{array} \end{matrix}$$

$$\begin{aligned} D^1[2,3] &= \min(D^0[2,3], \\ &\quad D^0[2,1] + D^0[1,3]) \\ &= \min(\infty, 7) \\ &= 7 \end{aligned}$$

$$\Pi = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{array}{|c|c|c|} \hline 0 & 0 & 0 \\ \hline 0 & 0 & 1 \\ \hline 0 & 0 & 0 \\ \hline \end{array} \end{matrix}$$

$$\begin{aligned} D^1[3,2] &= \min(D^0[3,2], \\ &\quad D^0[3,1] + D^0[1,2]) \\ &= \min(-3, \infty) \\ &= -3 \end{aligned}$$

$k = 2$ Vertices 1, 2 can be intermediate



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$$D^1 =$$

	1	2	3
1	0	4	5
2	2	0	7
3	∞	-3	0

$$D^2 =$$

	1	2	3
1	0	4	5
2	2	0	7
3	-1	-3	0

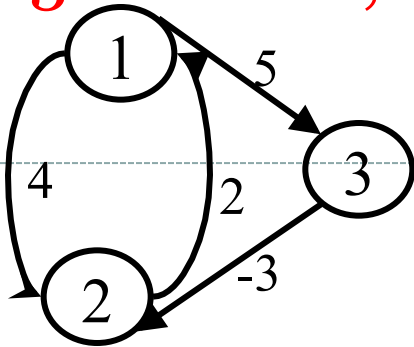
$$\begin{aligned} D^2[1,3] &= \min(D^1[1,3], \\ &\quad D^1[1,2] + D^1[2,3]) \\ &= \min(5, 4 + 7) \\ &= 5 \end{aligned}$$

$$\Pi =$$

	1	2	3
1	0	0	0
2	0	0	1
3	2	0	0

$$\begin{aligned} D^2[3,1] &= \min(D^1[3,1], \\ &\quad D^1[3,2] + D^1[2,1]) \\ &= \min(\infty, -3 + 2) \\ &= -1 \end{aligned}$$

$k = 3$ Vertices 1, 2, 3 can be intermediate



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$$D^2 =$$

	1	2	3
1	0	4	5
2	2	0	7
3	-1	-3	0

$$D^3 =$$

	1	2	3
1	0	2	5
2	2	0	7
3	-1	-3	0

$$\Pi =$$

	1	2	3
1	0	3	0
2	0	0	1
3	2	0	0

$$\begin{aligned} D^3[1,2] &= \min(D^2[1,2], \\ &\quad D^2[1,3] + D^2[3,2]) \\ &= \min(4, 5 + (-3)) \\ &= 2 \end{aligned}$$

$$\begin{aligned} D^3[2,1] &= \min(D^2[2,1], \\ &\quad D^2[2,3] + D^2[3,1]) \\ &= \min(2, 7 + (-1)) \\ &= 2 \end{aligned}$$

Floyd's Algorithm: Using 2 D matrices

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```
 $D \leftarrow W$     // initialize  $D$  array to  $W$  [ ]  
 $\Pi \leftarrow \emptyset$     // initialize  $\Pi$  array to [0]  
for  $k \leftarrow 1$  to  $n$   
    //Computing  $D'$  from  $D$   
    for  $i \leftarrow 1$  to  $n$   
        for  $j \leftarrow 1$  to  $n$   
            if ( $D[i, j] > D[i, k] + D[k, j]$ )  
                then  $D'[i, j] \leftarrow D[i, k] + D[k, j]$   
                     $\Pi[i, j] \leftarrow k$ ;  
            else  $D'[i, j] \leftarrow D[i, j]$   
Move  $D'$  to  $D$ .
```

Printing shortest path from q to r

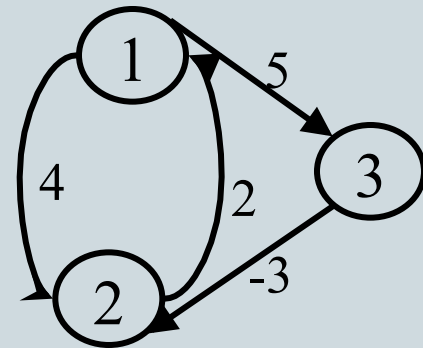
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```
path(index q, r)
  if ( $\Pi[q, r] \neq \emptyset$ )
    path(q,  $\Pi[q, r]$ )
    println( "v" +  $\Pi[q, r]$ )
    path( $\Pi[q, r]$ , r)
    return;
  //no intermediate nodes
  else return
```

Before calling path check $D[q, r] < \infty$, and
print node q, after the call to
path print node r

$\Pi =$

	1	2	3
1	0	3	0
2	0	0	1
3	2	0	0



Longest Common Subsequence (LCS)

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- A subsequence of a sequence/string S is obtained by deleting zero or more symbols from S . For example, the following are **some** subsequences of “president”: pred, sdn, predent. In other words, the letters of a subsequence of S appear in order in S , but they are not required to be consecutive.
- The longest common subsequence problem is to find a maximum length common subsequence between two sequences.
- Applications? Compare the DNA of two (or more) different organisms

DNA

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A strand of DNA consists of a string of molecules called **bases**, possible bases: adenine (A), guanine(G), cytosine(C), and thymine(T).

For example:

the DNA of one organism may be

$S_1 = \text{ACCGGTCGAGTGCGCGGAAGCCGGCCGAA}$, and

the DNA of another organism may be

$S_2 = \text{GTCGTTCGGAATGCCGTTGCTCTGTAAA}$.

One reason to compare two strands of DNA is to determine how “similar” the two strands are

LCS

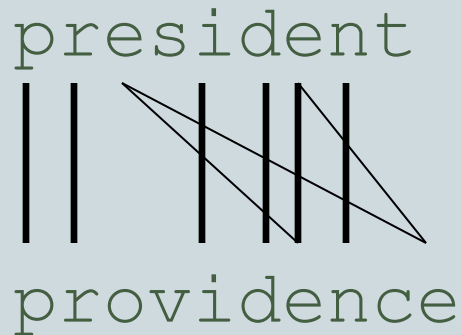
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Example1

Sequence 1: president

Sequence 2: providence

Its LCS is priden.



LCS

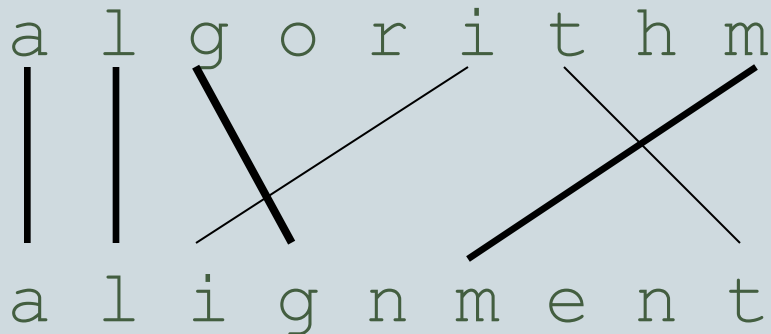
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Example2

Sequence 1: algorithm

Sequence 2: alignment

One of its LCS is algm.



How to compute LCS?

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- Let $A=x_1x_2...x_m$ and $B=y_1y_2...y_n$.
- $c[i, j]$: the length of an LCS between $x_1x_2...x_i$ and $y_1y_2...y_j$
- With proper initializations, $c[i, j]$ can be computed as follows.

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

Algorithm for LCS

LCS-LENGTH(X, Y)

```
1   $m = X.length$ 
2   $n = Y.length$ 
3  let  $b[1..m, 1..n]$  and  $c[0..m, 0..n]$  be new tables
4  for  $i = 1$  to  $m$ 
5       $c[i, 0] = 0$ 
6  for  $j = 0$  to  $n$ 
7       $c[0, j] = 0$ 
8  for  $i = 1$  to  $m$ 
9      for  $j = 1$  to  $n$ 
10         if  $x_i == y_j$ 
11              $c[i, j] = c[i - 1, j - 1] + 1$ 
12              $b[i, j] = \nwarrow$ 
13         elseif  $c[i - 1, j] \geq c[i, j - 1]$ 
14              $c[i, j] = c[i - 1, j]$ 
15              $b[i, j] = \uparrow$ 
16         else  $c[i, j] = c[i, j - 1]$ 
17              $b[i, j] = \leftarrow$ 
18  return  $c$  and  $b$ 
```

Running time: $\Theta(mn)$

Example of algorithm

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i	j	0	1	2	3	4	5	6	7	8	9	10
			<i>p</i>	<i>r</i>	<i>o</i>	<i>v</i>	<i>i</i>	<i>d</i>	<i>e</i>	<i>n</i>	<i>c</i>	<i>e</i>
0		0	0	0	0	0	0	0	0	0	0	0
1	<i>p</i>	0 ↘	1 ←	1 ←	1 ←	1 ←	1 ←	1 ←	1 ←	1 ←	1 ←	1 ←
2	<i>r</i>	0 ↑	1 ↘	2 ←	2 ←	2 ←	2 ←	2 ←	2 ←	2 ←	2 ←	2 ←
3	<i>e</i>	0 ↑	1 ↑	2 ↑	2 ↑	2 ↑	2 ↑	2 ↘	3 ←	3 ←	3 ←	3 ↘
4	<i>s</i>	0 ↑	1 ↑	2 ↑	2 ↑	2 ↑	2 ↑	2 ↑	3 ↑	3 ↑	3 ↑	3 ↑
5	<i>i</i>	0 ↑	1 ↑	2 ↑	2 ↑	2 ↘	3 ←	3 ↑	3 ↑	3 ↑	3 ↑	3 ↑
6	<i>d</i>	0 ↑	1 ↑	2 ↑	2 ↑	2 ↑	3 ↘	4 ←	4 ←	4 ←	4 ←	4 ←
7	<i>e</i>	0 ↑	1 ↑	2 ↑	2 ↑	2 ↑	3 ↑	4 ↘	5 ←	5 ←	5 ←	5 ↘
8	<i>n</i>	0 ↑	1 ↑	2 ↑	2 ↑	2 ↑	3 ↑	4 ↑	5 ↘	6 ←	6 ←	6 ←
9	<i>t</i>	0 ↑	1 ↑	2 ↑	2 ↑	2 ↑	3 ↑	4 ↑	5 ↑	6 ↑	6 ↑	6 ↑

Constructing a LCS

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PRINT-LCS(b, X, i, j)

1 **if** $i == 0$ or $j == 0$

2 **return**

3 **if** $b[i, j] == \nwarrow$

4 PRINT-LCS($b, X, i - 1, j - 1$)

5 print x_i

6 **elseif** $b[i, j] == \uparrow$

7 PRINT-LCS($b, X, i - 1, j$)

8 **else** PRINT-LCS($b, X, i, j - 1$)

Running time: $\Theta(m+n)$

Initial call PRINT-LCS($b, X, X.length, Y.length$)

Example of LCS

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i \ j	0	1	2	3	4	5	6	7	8	9	10
		<i>p</i>	<i>r</i>	<i>o</i>	<i>v</i>	<i>i</i>	<i>d</i>	<i>e</i>	<i>n</i>	<i>c</i>	<i>e</i>
0	0	0	0	0	0	0	0	0	0	0	0
1 <i>p</i>	0	1	1	1	1	1	1	1	1	1	1
2 <i>r</i>	0	1	2	2	2	2	2	2	2	2	2
3 <i>e</i>	0	1	2	2	2	2	3	3	3	3	3
4 <i>s</i>	0	1	2	2	2	2	3	3	3	3	3
5 <i>i</i>	0	1	2	2	2	3	3	3	3	3	3
6 <i>d</i>	0	1	2	2	2	3	4	4	4	4	4
7 <i>e</i>	0	1	2	2	2	3	4	5	5	5	5
8 <i>n</i>	0	1	2	2	2	3	4	5	6	6	6
9 <i>t</i>	0	1	2	2	2	3	4	5	6	6	6

Output: *priden*