



# String Distance and Dynamic Programming

---

## Lecture 5



# Life is similar

---

- Life is based on a repertoire of successful structural and interrelated building blocks which are passed around
- The vast majority of proteins are the result of a series of genetic duplications and subsequent modifications
- “Everything in life is so similar that the same genes that work in flies are the ones that work in humans” (Wieschaus, 1995)



# Comparison and analogy

---

- By identifying and comparing related objects we can distinguish variable and conserved features, and thereby determine what is crucial to structure and function
- Biological universality occurs at many levels of details, so we can compare not only the sequence data, but 3D shapes, chemical pathways, morphological features etc.



# Why compare biosequences

---

- The biological sequences encode and reflect higher-level molecular structures and mechanisms
- **In bimolecular sequences (DNA, RNA or protein), high sequence similarity usually implies significant structural and functional similarity**
- A tractable, though partly heuristic way to infer the structure and function of an unknown protein is to search for the similar known proteins at the sequence level



# Keep in mind

---

- There is not a one-to-one correspondence between similar sequences and similar structures or between sequences and functions:
  - Similar structures can be obtained from completely unrelated sequences
  - Very similar sequences can produce very different structures depending on the location of a change



# A shift to approximate pattern matching

---

- Approximate – means some errors are allowed in valid matches
- The shift is accompanied by a shift in technique: *dynamic programming*

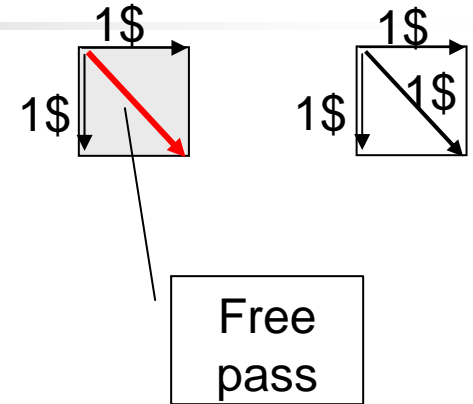
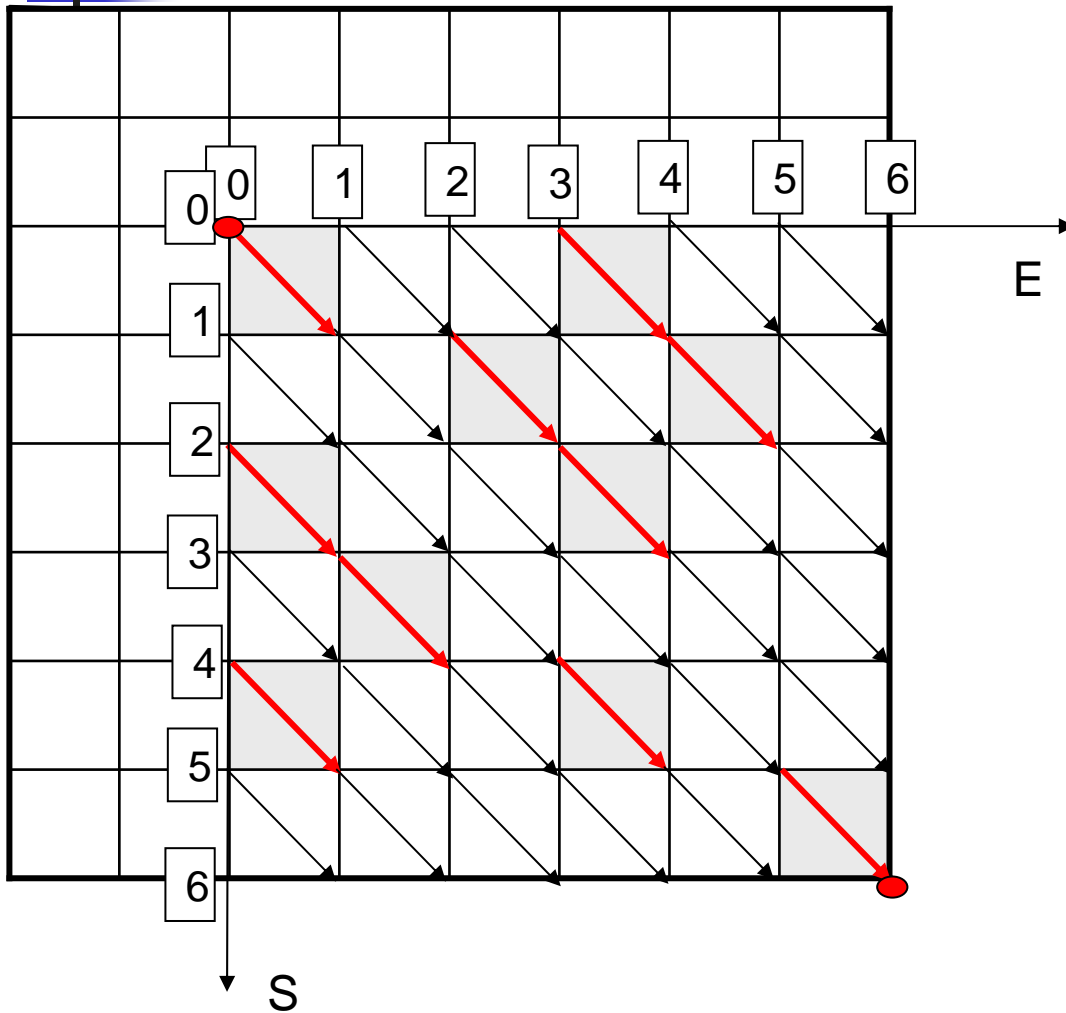


# Dynamic programming

---

The main tool in approximate  
pattern matching

# The cheapest path

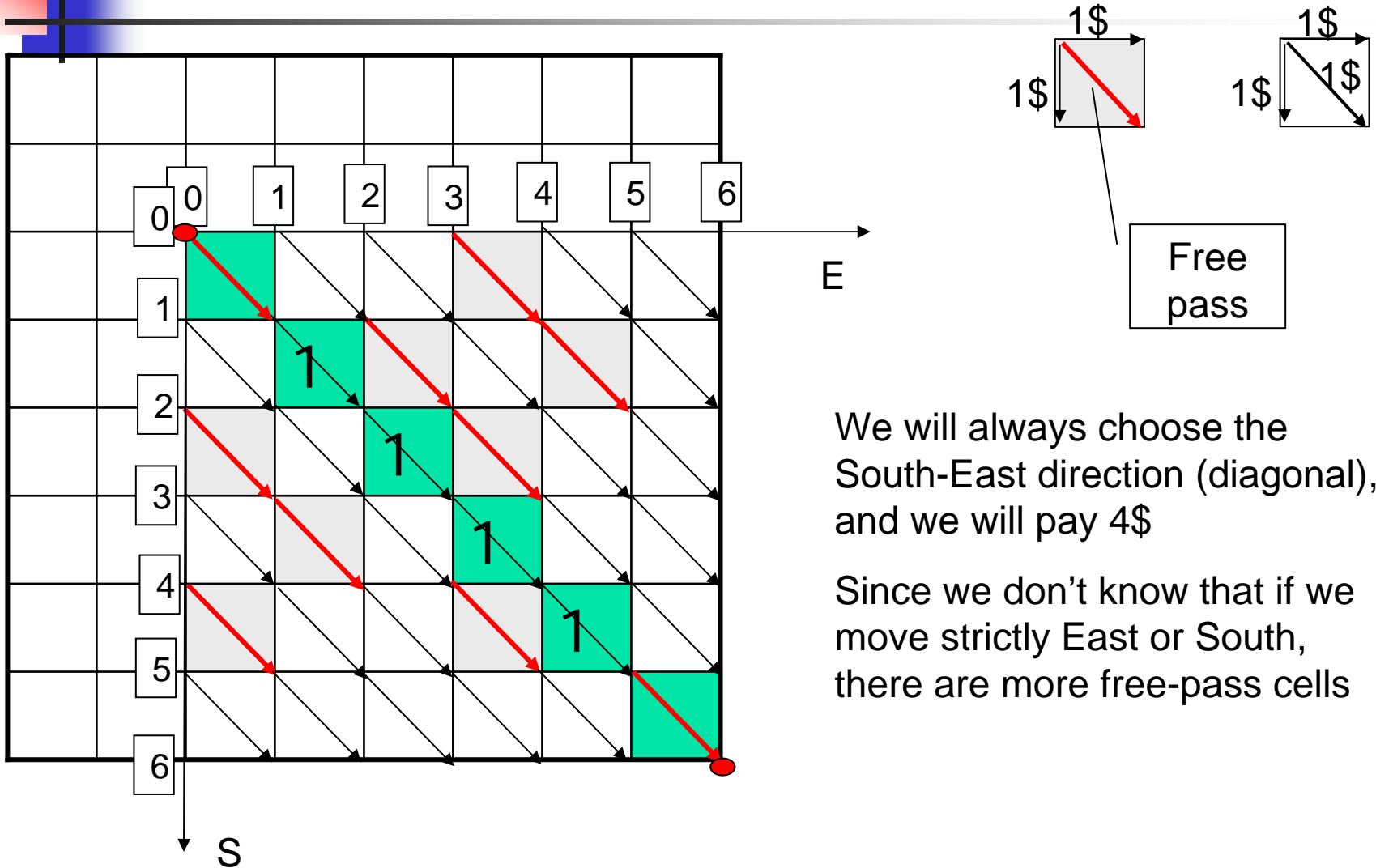


Problem:

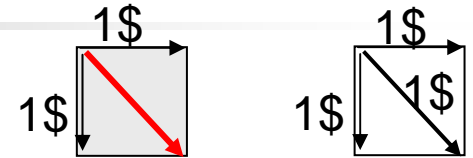
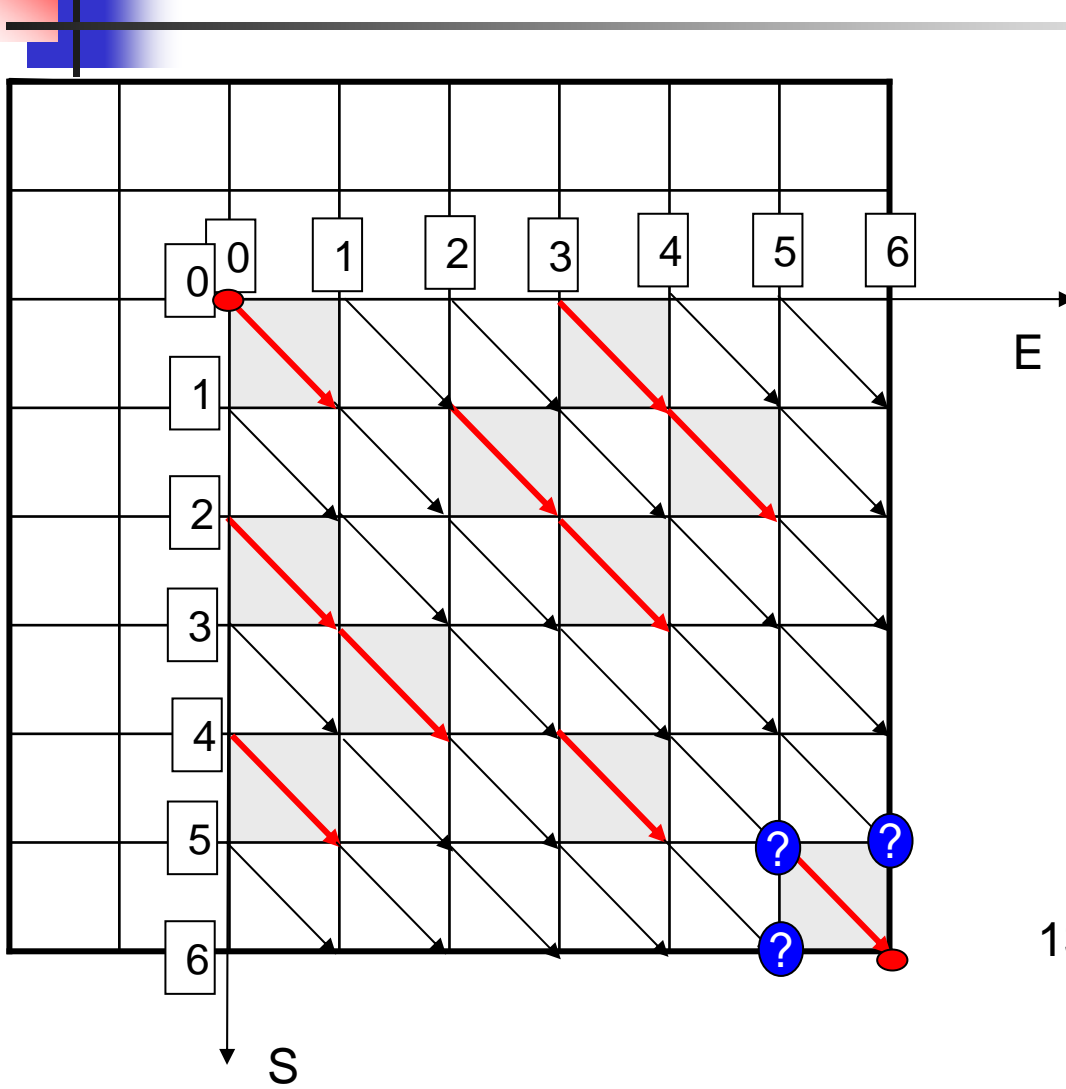
find the cheapest path  
from (0,0) to (6,6)



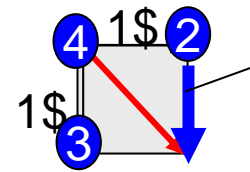
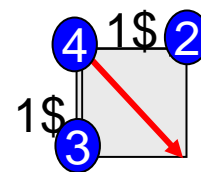
# The path without a map



# Sub-problems approach

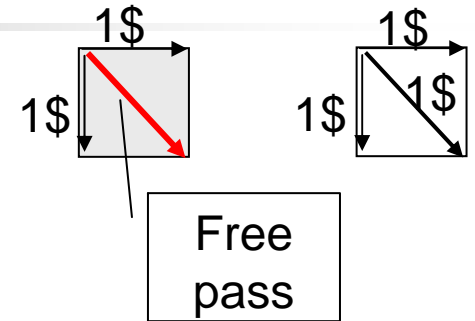
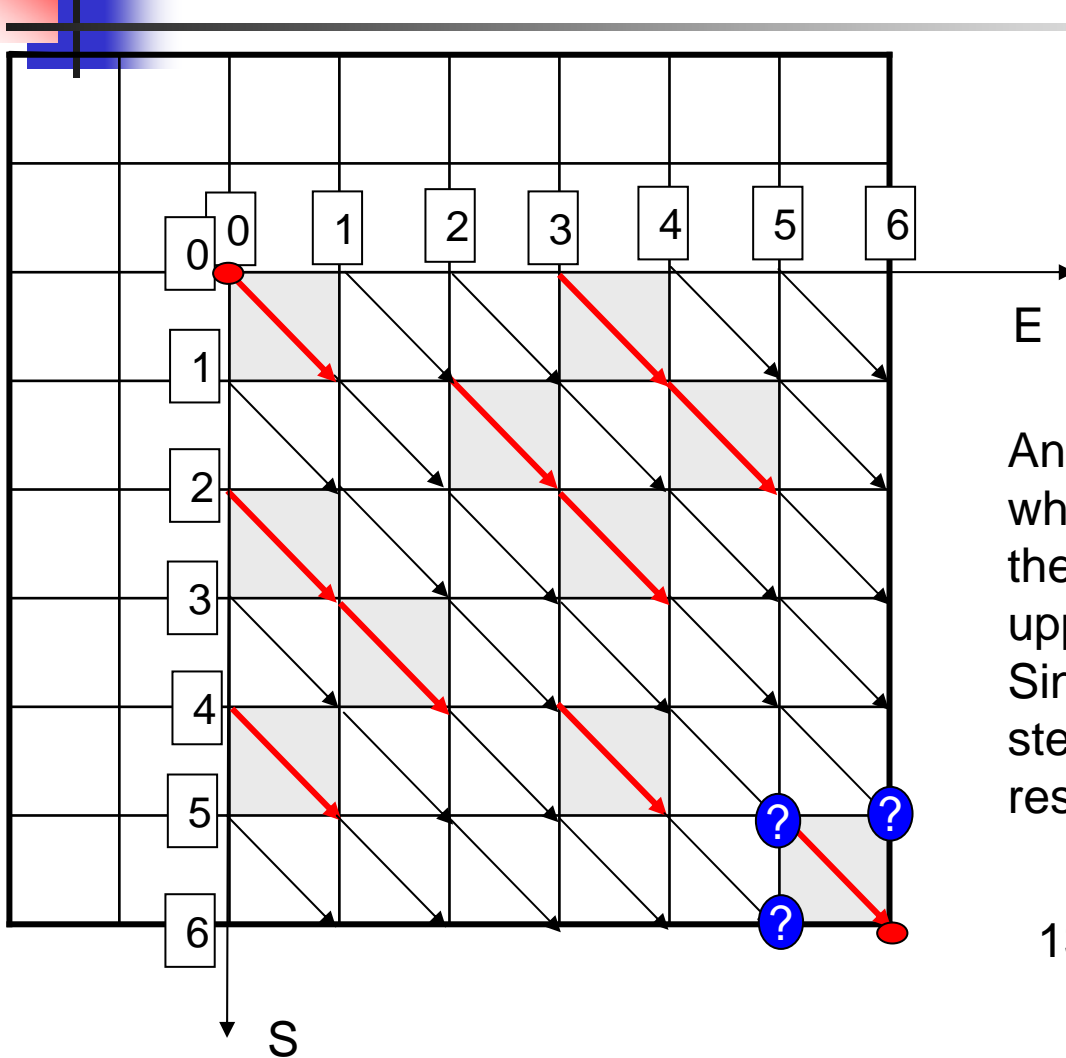


If we knew the cheapest paths  
 from (0,0) to (5,5)  
 from (0,0) to (6,5)  
 from (0,0) to (5,6)  
 we could choose the best  
 last step to the destination:  
 For example, if:

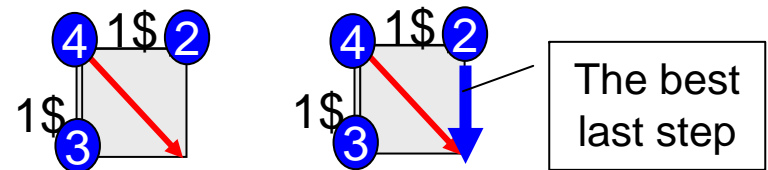


The best  
 last step

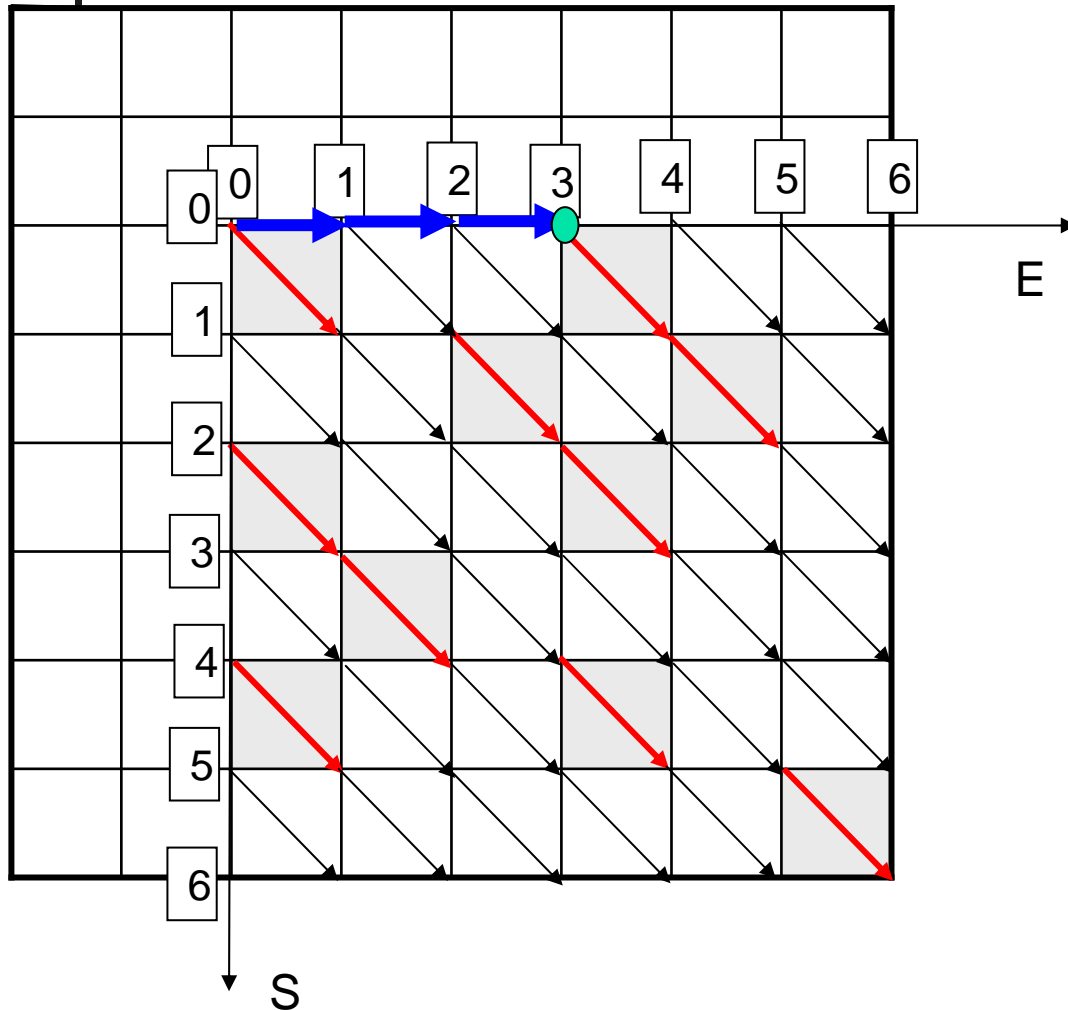
# The sub-problems approach



And this is true for any cell – what path to choose depends on the cheapest paths to the left, upper, and upper-left corner. Since we are choosing only 1 step, we can take the min of the result



# The recurrence relation – base condition



When  $i=0$ , there is no cheaper way of going from  $(0,0)$  to  $(0,j)$  than to pay  $j$  \$ - heading strictly to the right (East)

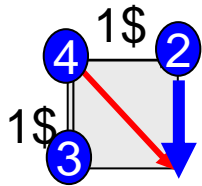
The same for  $j=0$ .

The base condition:

if  $i=0$  then  $COST(i,j)=j$

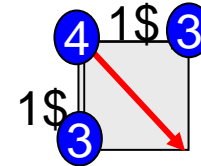
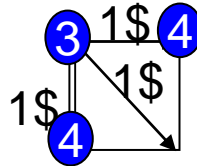
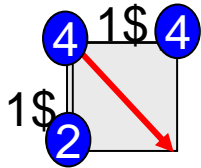
if  $j=0$  then  $COST(i,j)=i$

# The recurrence relation (for $i > 0$ and $j > 0$ )



$$\text{COST}(i,j) = \min \begin{cases} \text{COST}(i-1,j)+1 \\ \text{COST}(i,j-1)+1 \\ \text{COST}(i-1,j-1)+\text{DIAGONAL}(i,j) \end{cases}$$

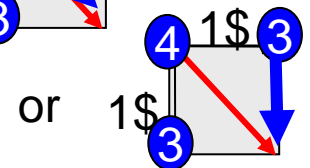
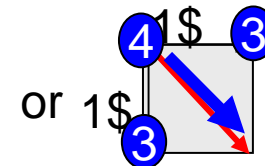
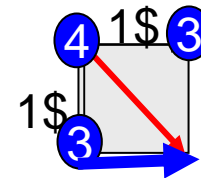
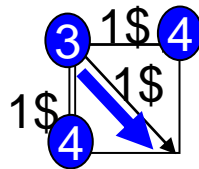
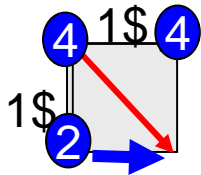
For each case, what is the best move?



# The recurrence relation

$$\text{COST}(i,j) = \min \begin{cases} \text{COST}(i-1,j)+1 \\ \text{COST}(i,j-1)+1 \\ \text{COST}(i-1,j-1)+\text{DIAGONAL}(i,j) \end{cases}$$

The best moves:



# The top-down (usual) recursion

$$\text{COST}(i,j)=\min \left\{ \begin{array}{l} \text{COST}(i-1,j)+1 \\ \text{COST}(i,j-1)+1 \\ \text{COST}(i-1,j-1)+\text{DIAGONAL}(i,j) \end{array} \right.$$

**algorithm cheapestCost** ( array *diagonalCost*, *N*, *M* )

**return** *cost* ( *N*, *M* )

**algorithm cost** ( *i*, *j* )

**if** *i*=0 **then**

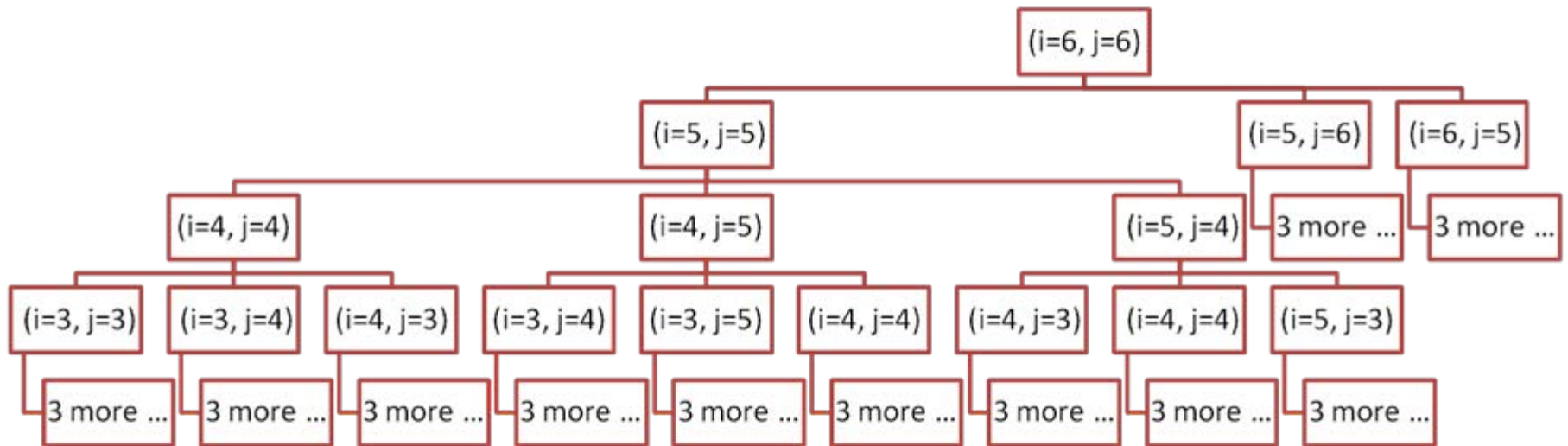
**return** *j*

**if** *j*=0 **then**

**return** *i*

**return** **min** ( *cost* ( *i*-1, *j* ) +1, *cost* ( *i*, *j*-1 ) +1, *cost* ( *i*-1, *j*-1 ) +*diagonalCost* [*i*] [*j*] ) )

# The recursion tree

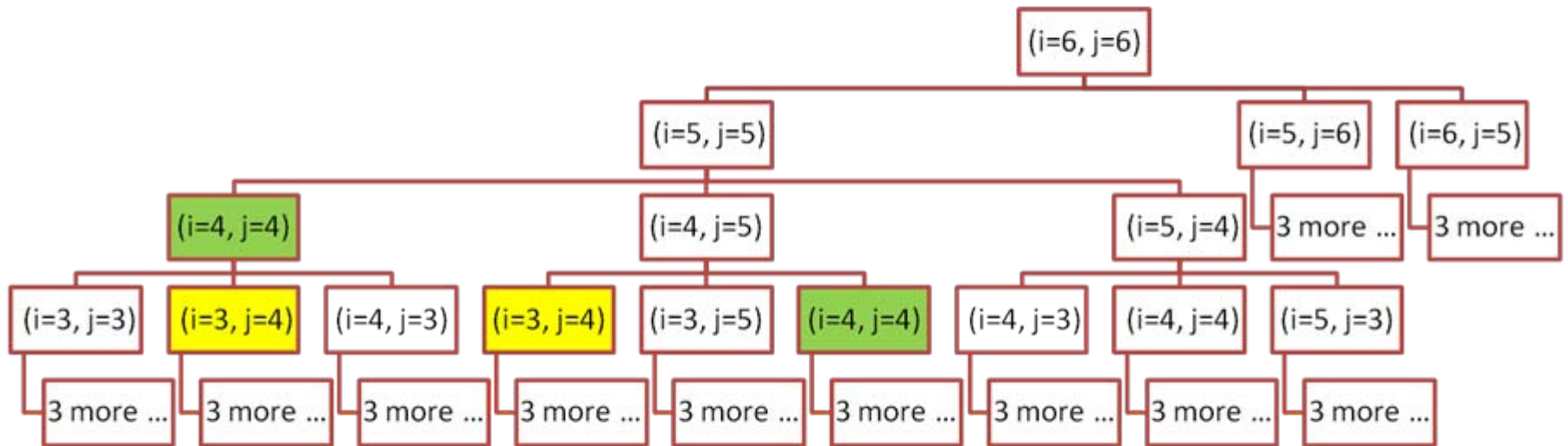


$O(3^N)$  ?

But there are only  $N \cdot M$   
different combinations



# The recursion tree



$O(3^N)$  ?

We call the recursive function multiple times with the same parameters



# Dynamic programming steps

---

- The recurrence relation
- The bottom-up computation
- The traceback



# Dynamic programming I

---

- **The recurrence relation**
  - The bottom-up computation
  - The traceback



# The recurrence relation

---

The base condition:

if $i=0$ then $COST(i,j)=j$ if $j=0$ then $COST(i,j)=i$
--

The main relation ( for  $i>0$  and  $j>0$ )

$COST(i,j)=\min$	$\left\{ \begin{array}{l} COST(i-1,j)+1 \\ COST(i,j-1)+1 \\ COST(i-1,j-1)+DIAGONAL(i,j) \end{array} \right.$
------------------	--



# Dynamic programming II

---

- The recurrence relation
- **The bottom-up computation**
- The traceback

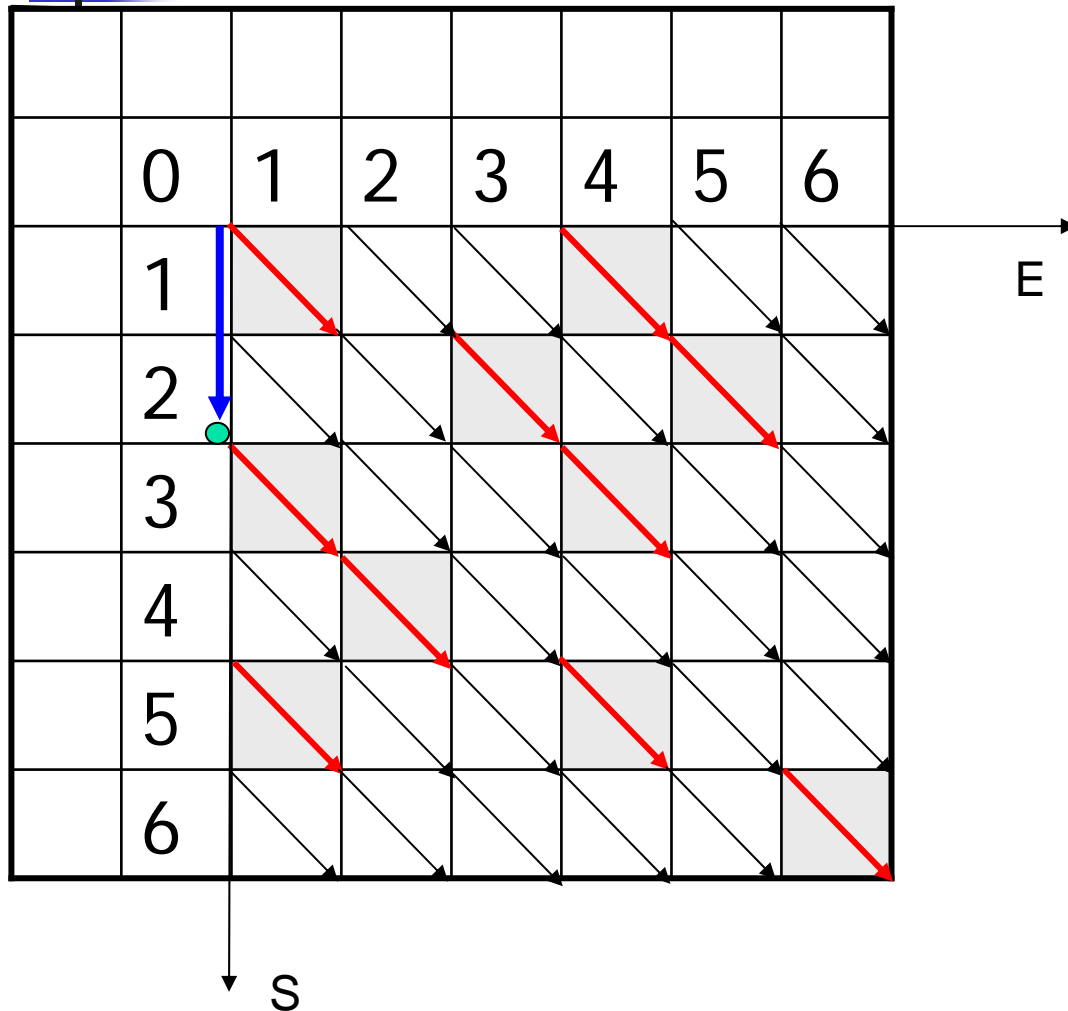


# The bottom-up computation

---

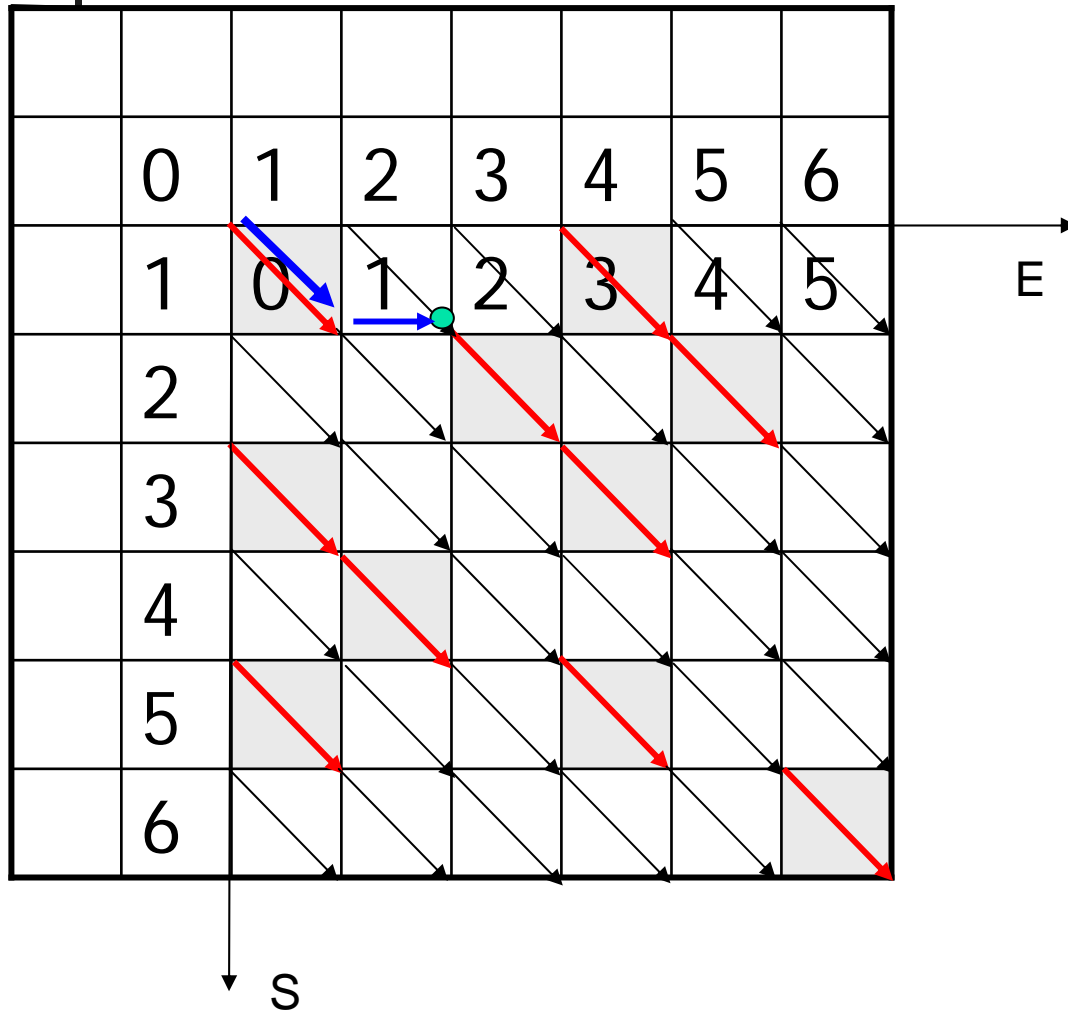
- Fill in the best values for each cell of the  $N \times M$  table starting from the lowest values
- First, compute the basic values of recursion – for  $i=0$  and for  $j=0$
- Apply recursion relation for computing the value of each cell from the lowest numbers of  $i$  and  $j$  to the largest
- At the end, we will have the cost of the best path in the cell  $(N, M)$

# Fill values for $i=0$ and for $j=0$ (the base recursion condition)



There is no cheaper way of going to the point (2,0) than paying 2 \$

# Fill values for $i=1$ (from left to right)



Cell(1,2)=1

since the  
cheapest  
possible way is  
to continue the  
free path  
through the cell  
(1,1)



# Fill in the entire table (left-to-right top-down)

	0	1	2	3	4	5	6
1	0	1	2	3	4	5	
2	1	1	1	2	3	4	
3	2	2	2	1	2	3	
4	3	2	3	2	2	3	
5	4	3	3	3	3	3	
6	5	4	4	4	4	3	

The cheapest possible path costs 3\$

But what is this path?

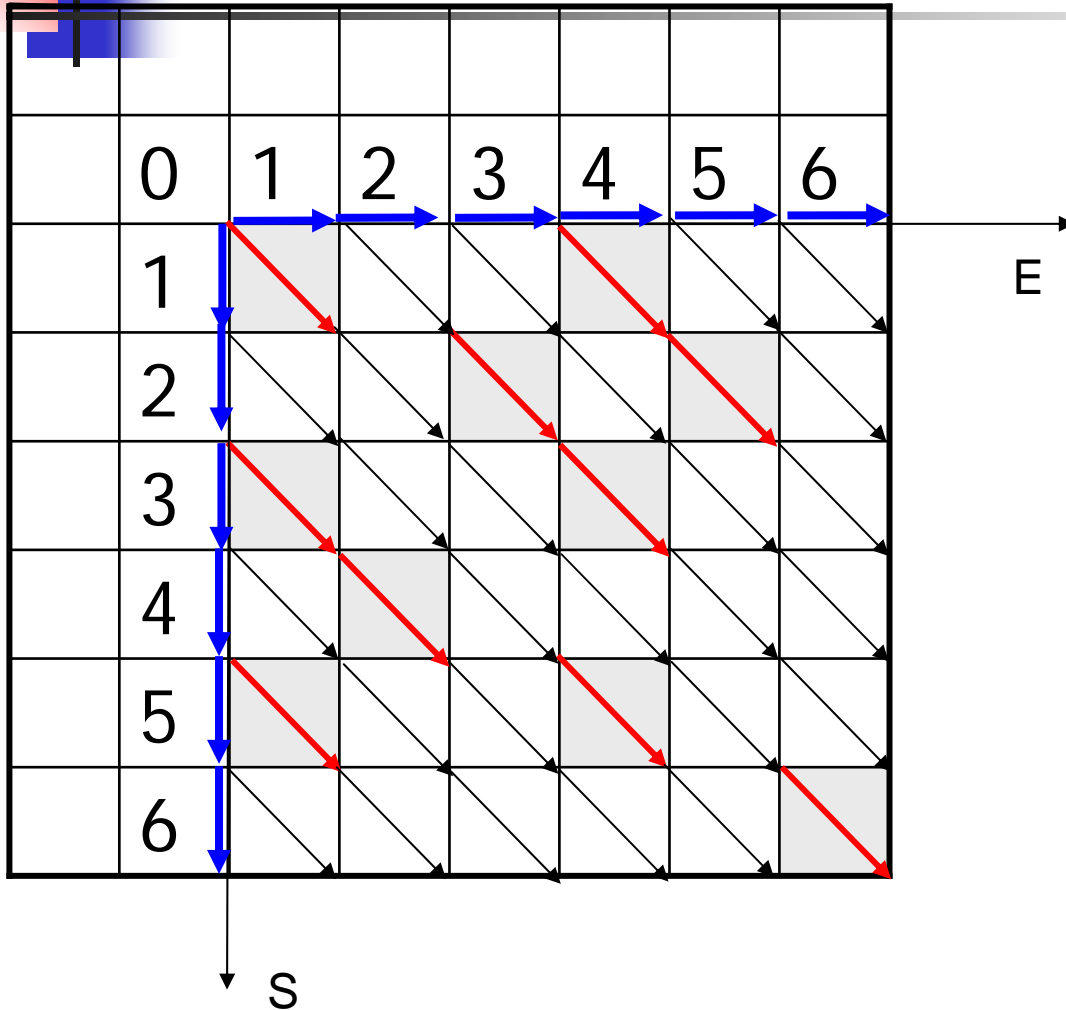


# Dynamic programming III

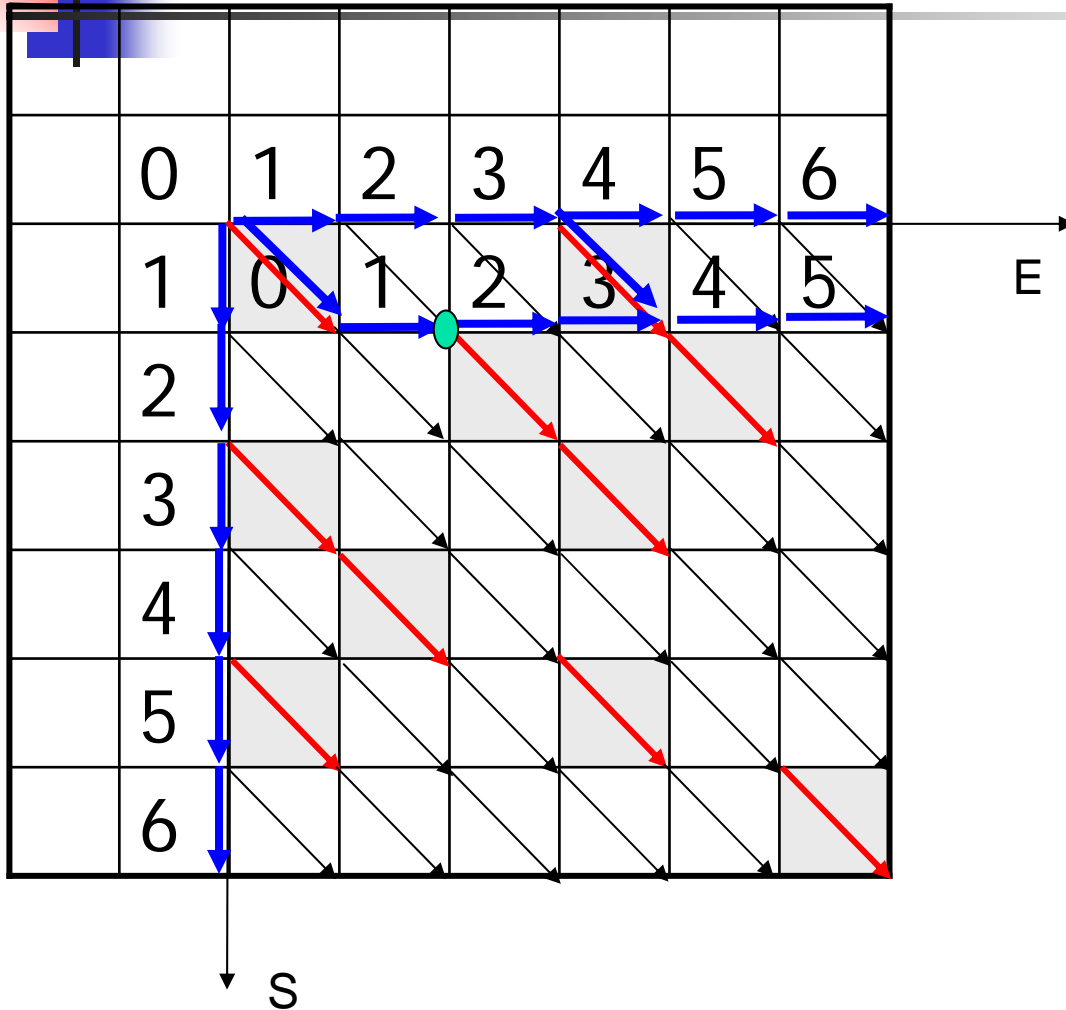
---

- The recurrence relation
- The bottom-up computation
- **The traceback**

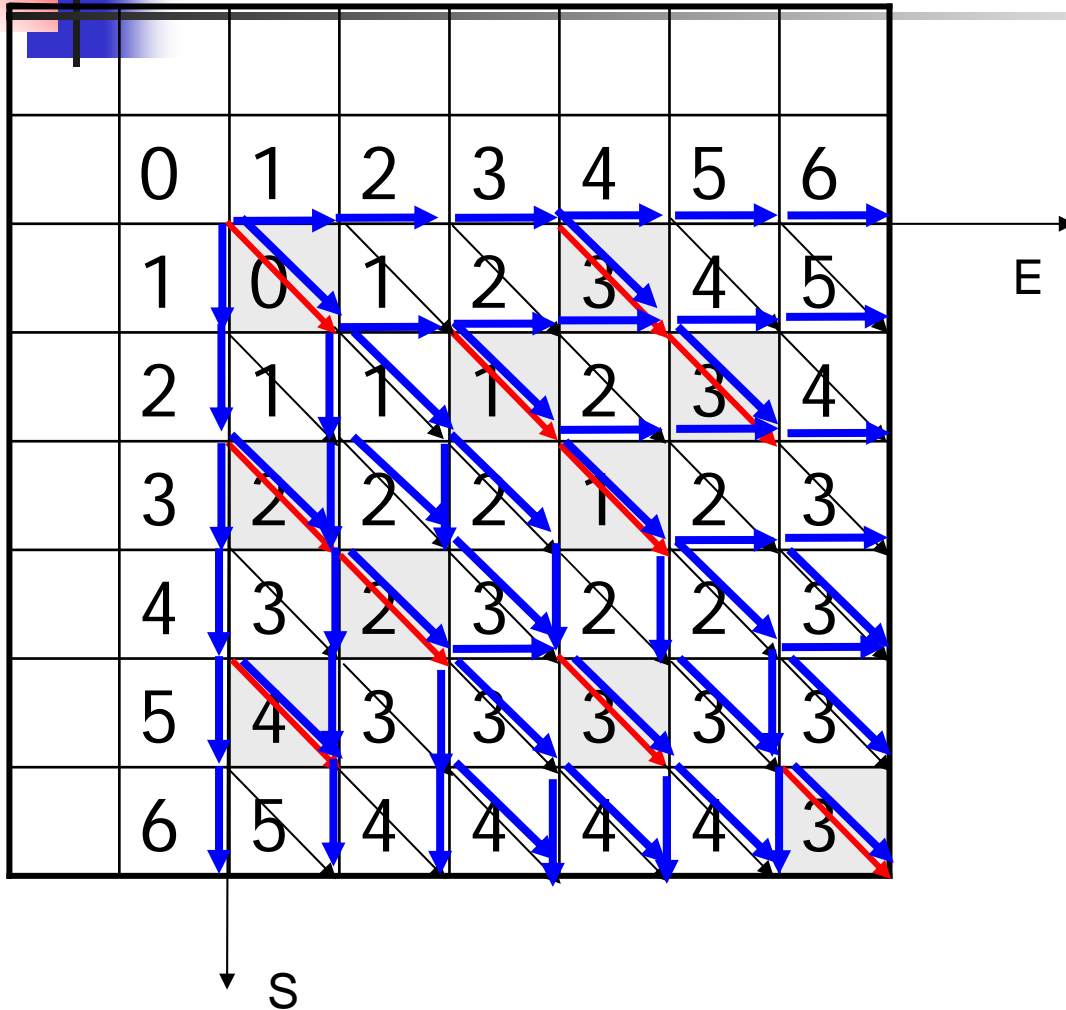
# Keeping track of the source



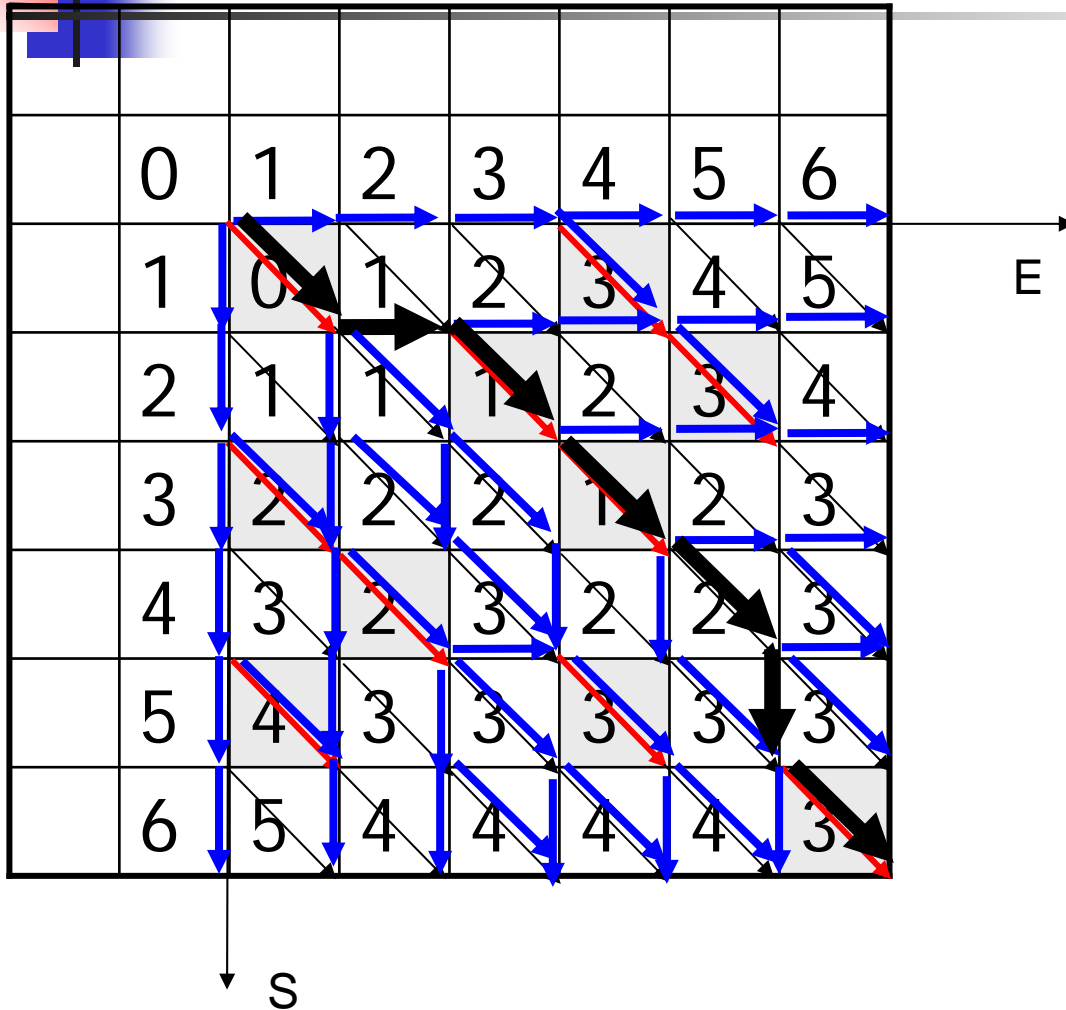
# Keeping track of the source

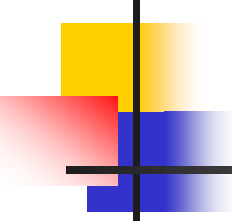


# Keeping track of the source



Trace back –  
 how did we get the path with the cost 3





# Dynamic programming with electronic tables. Cost

---

- Build the input table – the cost of passing through any cell by diagonal
- Create the distance table, fill the first row and the first column according to the basic recursion
- Insert the recursion formula in cell [1][1]:
  - $C19 = \text{MIN}(B18 + C3, B19 + 1, C18 + 1)$
- Spread the formula to the rest of the table by drag-and-release
- Read the cost of the cheapest path in cell [N][M] – the last cell of the cost table

# Dynamic programming with electronic tables. Cost

The cost of passing through the corresponding cell of the input table

$$C19 = \text{MIN}(B18 + C3, B19 + 1, C18 + 1)$$

Current cell:  
 $i=C, j=19$

$i-1=B,$   
 $j-1=18$

$i-1=B,$   
 $j=19$

$i=C,$   
 $j-1=18$





# Dynamic programming with electronic tables. Forward path

## Excel code

```
C35=  
IF(B18+C3<B19+1,  
    IF(B18+C3<C18+1,  
        "DownRight",  
        "Right"),  
    "Down")
```

```
IF(B18+C3<B19+1)  
    IF(B18+C3<C18+1)  
        C35="DownRight"  
    ELSE  
        C35="Right"  
ELSE  
    C35="Down"
```

Shows one of the possible paths to obtain the smallest cost for a path from (0,0) to the current cell



# Dynamic programming with electronic tables. Backward path

## Excel code

```
C49=  
IF(C35="Down",  
    "Up",  
    IF(C35="Right",  
        "Left",  
        "UpLeft"))
```

```
IF(C35 = "Down")  
    C49 = "Up"  
ELSE  
    IF(C35 = "Right")  
        C49 = "Left"  
    ELSE  
        C49 = "UpLeft"
```

Replacing by the opposite direction – from the destination cell to the source cell

# Dynamic programming with electronic tables. Traceback

## Excel code

```
B60=  
IF(AND(C61="X",C49="UpLeft"),  
    "X",  
IF(AND(C60="X",C48="Left"),  
    "X",  
IF(AND(B61="X",B49="Up"),  
    "X",  
    "-"))))
```

```
IF( C61="X"AND C49="UpLeft")  
    B60="X"  
ELSE IF( C60="X" AND C48="Left")  
    B60="X"  
ELSE IF( B61="X" AND B49="Up")  
    B60="X"  
ELSE  
    B60="-"
```

By placing X in the destination cell, this code reconstructs the path which gave the total minimum cost: cell is marked X if the path went through this cell, otherwise it is marked -.



Alternative: write the program  
(add the traceback and the output of the path)

**Input:** array *diagonalCost* ( $N \times M$ )  
**allocate array** *DPTable* ( $N \times M$ )

*algorithm* *getCheapestCost*( )  
    *fillDPTable*( )  
    return *DPTable* [ $N$ ] [ $M$ ]

*algorithm* *fillDPTable*( )  
    *DPTable* [0][0]:=0  
    for *i* from 1 to  $N$ :  
        *DPTable* [ $i$ ][0]:=*i*  
    for *j* from 1 to  $M$ :  
        *DPTable* [0][ $j$ ]:=*j*  
    for *i* from 1 to  $N$ :  
        for *j* from 1 to  $M$ :  
            *DPTable* [ $i$ ][ $j$ ]:=min (*DPTable* [ $i-1$ ][ $j-1$ ]+ *diagonalCost* [ $i$ ][ $j$ ],  
                                  *DPTable* [ $i-1$ ][ $j$ ]+1, *DPTable* [ $i$ ][ $j-1$ ]+ 1)



# Complexity of the DP algorithm

---

- 2 nested loops:  $O(NM)$



# Edit distance

---

String dissimilarity

# Edit Operations

- We can transform the second string S2 into the first string S1 by applying a sequence of edit operations on S2 :
  - Deleting 1 symbol
  - Inserting 1 symbol
  - Replacing 1 symbol

S1	a	c	t			a	t	g
S2	a		t	a	c	a		g

Insert c

Delete a, c

Insert t

In total, 4 edit operations

# String alignment

- An *alignment* of 2 strings is obtained by first inserting spaces (gaps), either into or at the end of both strings, and then placing the 2 resulting strings one above the other, so that every character or space in either string is opposite a single character or space in the other string

S1	a	c	t	-	-	a	t	g
S2	a	-	t	a	c	a	-	g

alignment

4 gaps,  
no mismatches



# Edit distance

- The *edit distance* between two strings is defined as the minimum number of edit operations needed to transform one string into another

S1	a	c	t	a	t		g
S2	a		t	a	c	a	g

Insert c

Replace c  
by t

Delete a

In total, 3 edit operations

# Optimal alignment

- An optimal alignment is obtained from the calculation of the edit distance

S1	a	c	t	a	t	-	g
S2	a	-	t	a	c	a	g

Optimal alignment

2 gaps,

1 mismatch

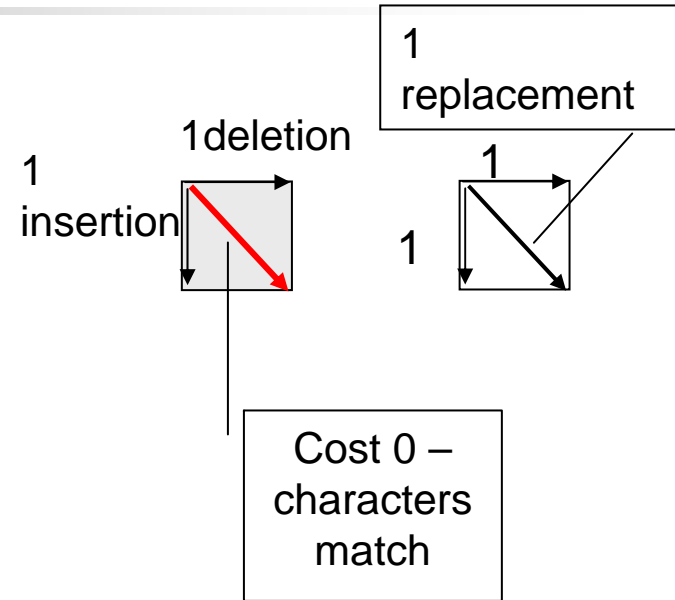
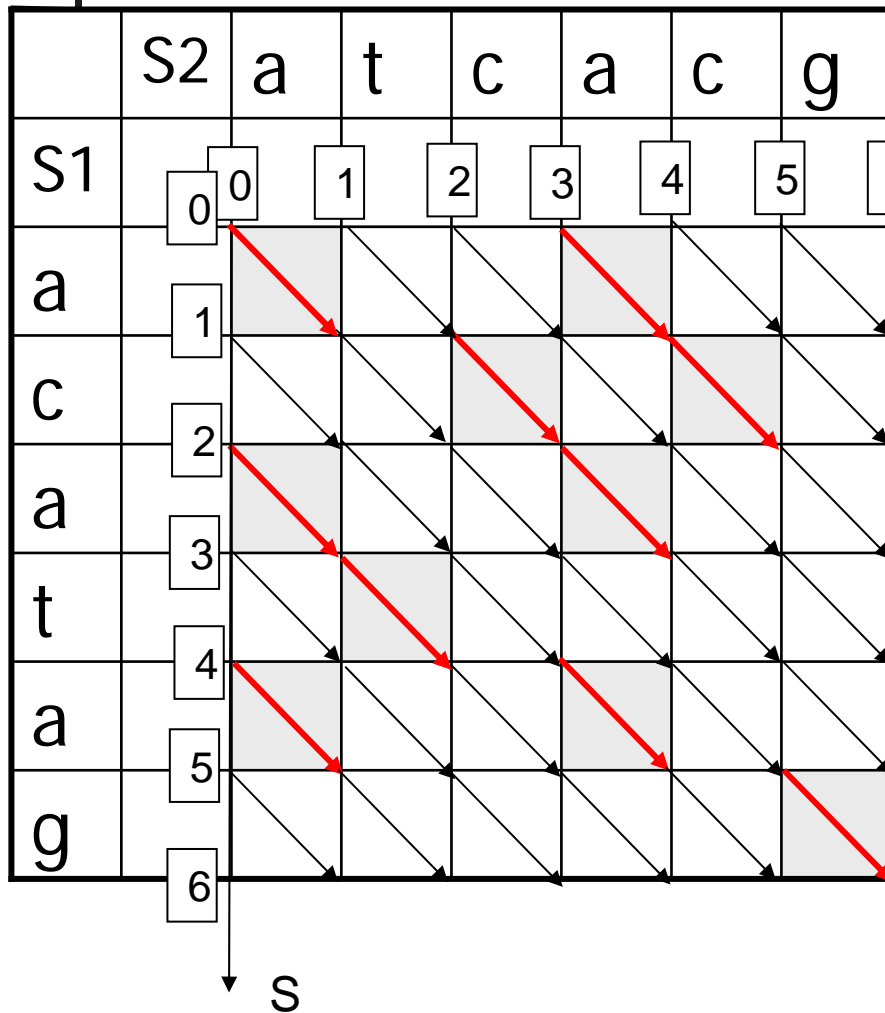


# The edit distance problem

---

- Compute the edit distance between two strings along with a sequence of the operations which describe the transformation

# Analogy with the cheapest path





# The dynamic programming solution to the edit distance problem

---

- Trivially follows from the solution for the cheapest path:
  - If we moved strictly down in the grid, we inserted 1 symbol into S2
  - If we moved strictly to the right, we deleted 1 symbol from S2
  - If we moved by diagonal of cost 0, we matched the corresponding characters
  - If we moved by diagonal of cost 1, we replaced one symbol in S2 with the corresponding symbol in S1