



# **Chapter 3**

# **Dynamic Programming**

**Algorithm Theory**  
**WS 2015/16**

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# Edit Distance

**Given:** Two strings  $A = \underline{a_1 a_2 \dots a_m}$  and  $B = \underline{b_1 b_2 \dots b_n}$

**Goal:** Determine the minimum number  $D(A, B)$  of edit operations required to transform  $A$  into  $B$

## Edit operations:

- a) **Replace** a character from string  $A$  by a character from  $B$
- b) **Delete** a character from string  $A$
- c) **Insert** a character from string  $B$  into  $A$

$\downarrow$ 
  
 m a ~~e~~ t h e m - - a t i c ~~x~~ a n
   
 m u l t i p l i c a t i o e - n

# Edit Distance – Cost Model $c(a, a) = 0$

- Cost for **replacing** character  $a$  by  $b$ :  $c(\underline{a}, \underline{b}) \geq 0$
- Capture insert, delete by allowing  $a = \varepsilon$  or  $b = \varepsilon$ :
  - Cost for **deleting** character  $a$ :  $c(\underline{a}, \underline{\varepsilon})$
  - Cost for **inserting** character  $b$ :  $c(\underline{\varepsilon}, \underline{b})$

- **Triangle inequality:**

$$c(a, c) \leq c(a, b) + c(b, c)$$

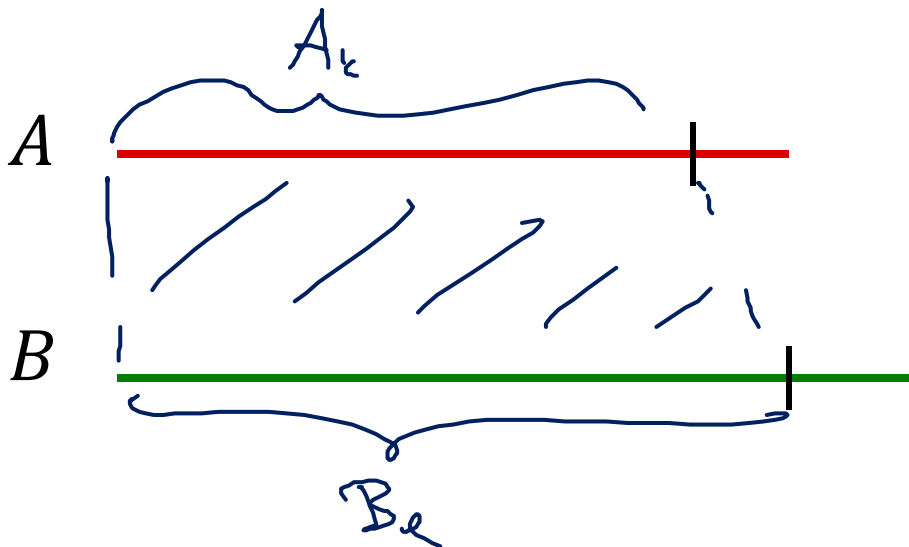
→ each character is changed at most once!

- **Unit cost model:**  $c(a, b) = \begin{cases} 1, & \text{if } a \neq b \\ 0, & \text{if } a = b \end{cases}$

# Computation of the Edit Distance

Let  $A_k := \underline{a_1 \dots a_k}$ ,  $B_\ell := \underline{b_1 \dots b_\ell}$ , and

$$\underline{D_{k,\ell}} := \underline{D(A_k, B_\ell)}$$



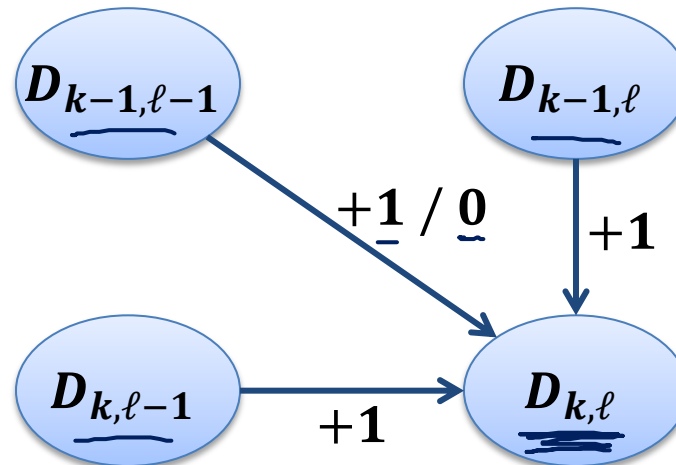
# Computing the Edit Distance

- Recurrence relation (for  $k, \ell \geq 1$ )

$$\underline{D}_{k,\ell} = \min \left\{ \begin{array}{l} \underline{D}_{k-1,\ell-1} + c(\underline{a_k}, \underline{b_\ell}) \\ \underline{D}_{k-1,\ell} + c(\underline{a_k}, \underline{\varepsilon}) \\ \underline{D}_{k,\ell-1} + c(\underline{\varepsilon}, \underline{b_\ell}) \end{array} \right\} = \min \left\{ \begin{array}{l} D_{k-1,\ell-1} + \underline{1} / \underline{0} \\ D_{k-1,\ell} + \underline{1} \\ D_{k,\ell-1} + \underline{1} \end{array} \right\}$$

unit cost model

- Need to compute  $D_{i,j}$  for all  $0 \leq i \leq k, 0 \leq j \leq \ell$ :



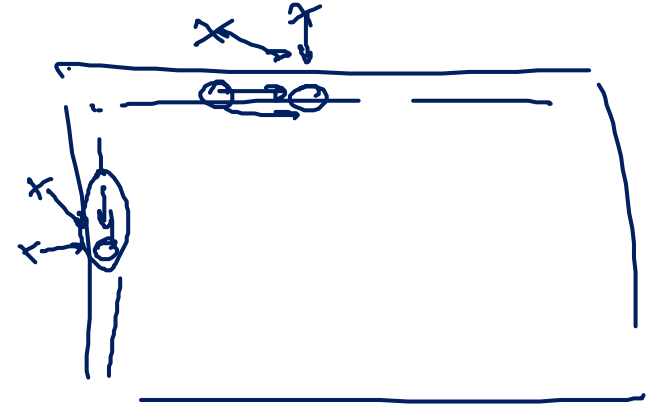
# Recurrence Relation for the Edit Distance

Base cases:

$$\underline{D_{0,0}} = D(\underline{\varepsilon}, \underline{\varepsilon}) = \underline{0}$$

$$D_{0,j} = D(\underline{\varepsilon}, \underline{B_j}) = \underline{D_{0,j-1}} + \underline{c(\varepsilon, b_j)}$$

$$D_{i,0} = D(\underline{A_i}, \underline{\varepsilon}) = \underline{D_{i-1,0}} + \underline{c(a_i, \varepsilon)}$$

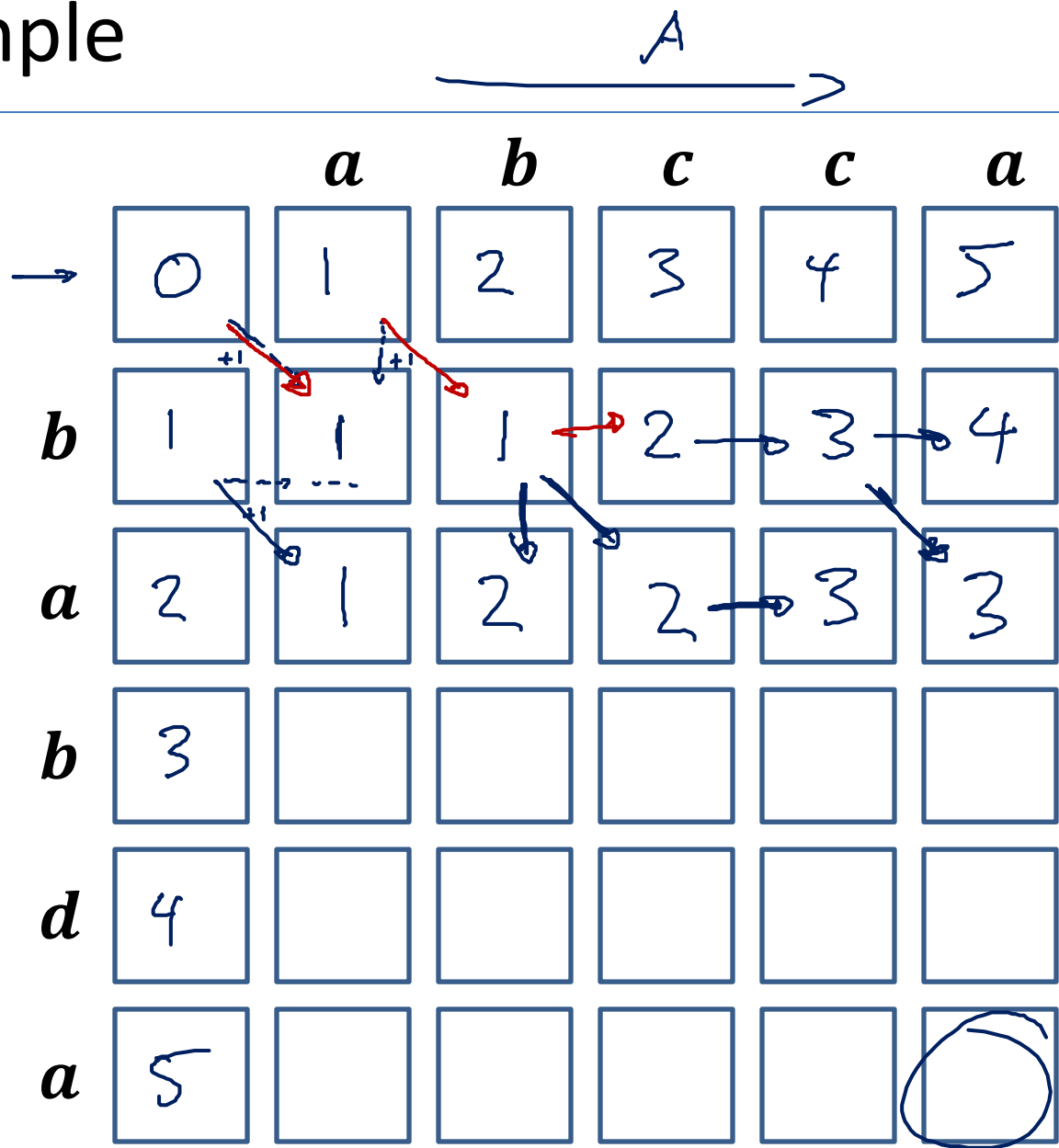


Recurrence relation:

$$D_{i,j} = \min \left\{ \begin{array}{l} D_{k-1,\ell-1} + c(a_k, b_\ell) \\ D_{k-1,\ell} + c(a_k, \varepsilon) \\ D_{k,\ell-1} + c(\varepsilon, b_\ell) \end{array} \right\}$$

# Example

B



# Computing the Edit Operations

**Algorithm** *Edit-Operations*( $i, j$ )

**Input:** matrix  $D$  (already computed)

**Output:** list of edit operations

- 1 **if**  $i = 0$  **and**  $j = 0$  **then return** empty list
- 2 **if**  $i \neq 0$  **and**  $D[i, j] = D[i - 1, j] + 1$  **then**
- 3     **return** *Edit-Operations*( $i - 1, j$ )  $\circ$  „delete  $a_i$ “
- 4 **else if**  $j \neq 0$  **and**  $D[i, j] = D[i, j - 1] + 1$  **then**
- 5     **return** *Edit-Operations*( $i, j - 1$ )  $\circ$  „insert  $b_j$ “
- 6 **else** //  $D[i, j] = D[i - 1, j - 1] + c(a_i, b_j)$
- 7     **if**  $a_i = b_i$  **then return** *Edit-Operations*( $i - 1, j - 1$ )
- 8     **else return** *Edit-Operations*( $i - 1, j - 1$ )  $\circ$  „replace  $a_i$  by  $b_j$ “

**Initial call:** *Edit-Operations*( $m, n$ )



# Edit Operations

		<i>a</i>	<i>b</i>	<i>c</i>	<i>c</i>	<u><i>a</i></u>
	0	1	2	3	4	5
<i>b</i>	1	1	1	2	3	4
<i>a</i>	2	1	2	2	3	3
<i>b</i>	3	2	1	2	3	4
<i>d</i>	4	3	2	2	3	4
<u><i>a</i></u>	5	4	3	3	3	3

*a* *b* *c* *c* *a*  
*b* *a* *b* *d* *a*

# Edit Distance: Summary

- Edit distance between two strings of length  $m$  and  $n$  can be computed in  $O(\underline{mn})$  time.
- Obtain the edit operations:
  - for each cell, store which rule(s) apply to fill the cell
  - track path backwards from cell  $(\underline{m}, n)$
  - can also be used to get all optimal “alignments”
- Unit cost model:
  - interesting special case
  - each edit operation costs 1

# Approximate String Matching

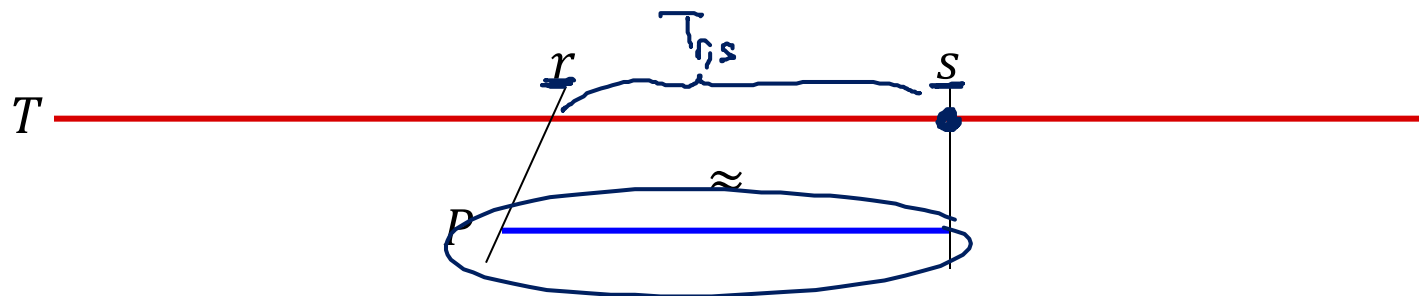
$m \ll n$



**Given:** strings  $T = \underline{t_1 t_2 \dots t_n}$  (text) and  $P = \underline{p_1 p_2 \dots p_m}$  (pattern).

**Goal:** Find an interval  $[r, s]$ ,  $1 \leq r \leq s \leq n$  such that the sub-string  $\underline{T_{r,s}} := t_r \dots t_s$  is the one with highest similarity to the pattern  $P$ :

$$\arg \min_{1 \leq r \leq s \leq n} \underline{D(T_{r,s}, P)}$$



# Approximate String Matching

## Naive Solution:

**for all**  $1 \leq r \leq s \leq n$  **do**

    compute  $D(T_{r,s}, P)$

choose the minimum

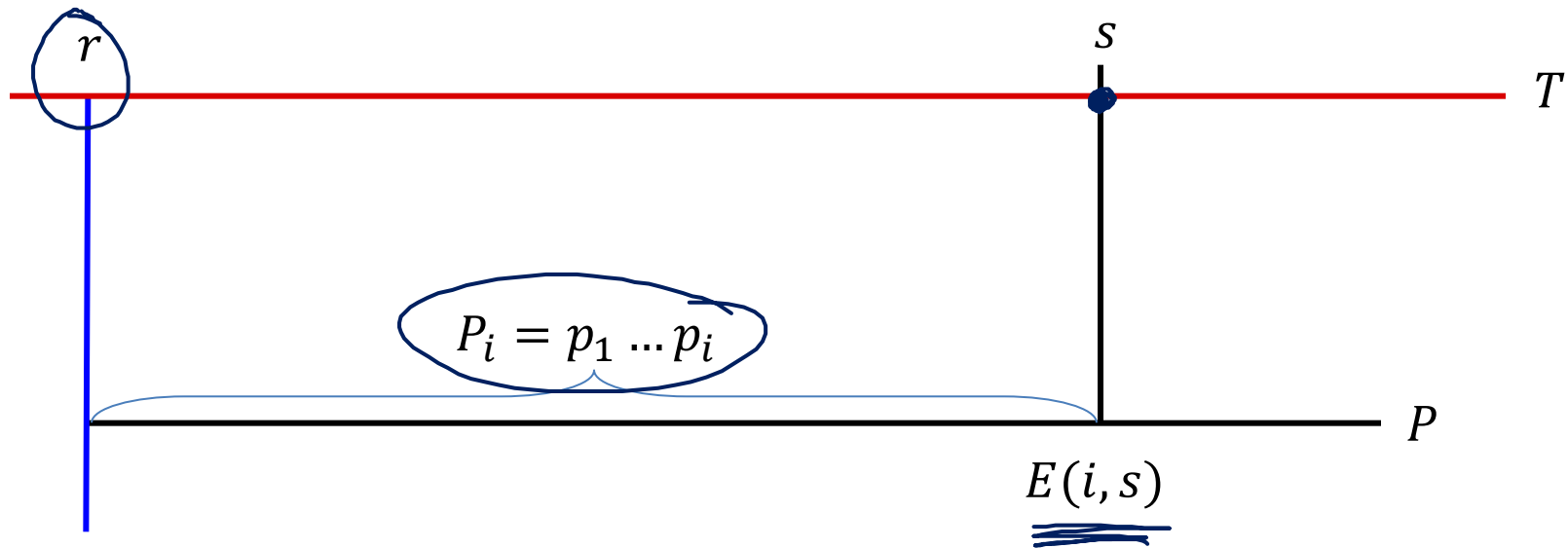
time:  $O(n^3 \cdot m)$

$O((s-r) \cdot m) = O(n \cdot m)$

# Approximate String Matching

A related problem:

- For each position  $s$  in the text and each position  $i$  in the pattern compute the minimum edit distance  $E(i, s)$  between  $P_i = \underline{p_1 \dots p_i}$  and any substring  $T_{r,s}$  of  $T$  that ends at position  $s$ .

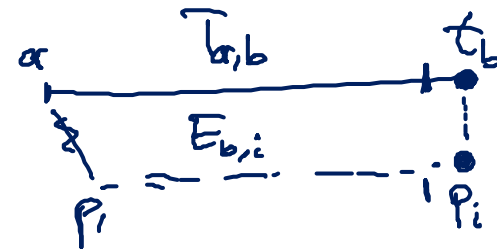


# Approximate String Matching

Three ways of ending optimal alignment between  $T_b$  and  $P_i$ :

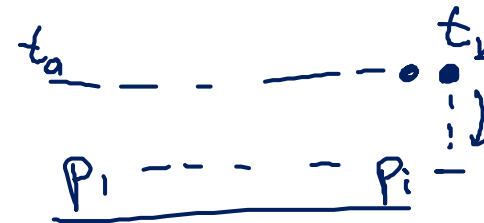
1.  $t_b$  is replaced by  $p_i$ :

$$E_{b,i} = \underline{E_{b-1,i-1}} + \underline{c(t_b, p_i)}$$



2.  $t_b$  is deleted:

$$E_{b,i} = \underline{E_{b-1,i}} + c(t_b, \varepsilon)$$



3.  $p_i$  is inserted:

$$E_{b,i} = \underline{E_{b,i-1}} + c(\varepsilon, p_i)$$



# Approximate String Matching

Recurrence relation (unit cost model):

$$E_{b,i} = \min \begin{cases} E_{b-1,i-1} + 1 \\ E_{b-1,i} + 1 \\ E_{b,i-1} + 1 \end{cases}$$

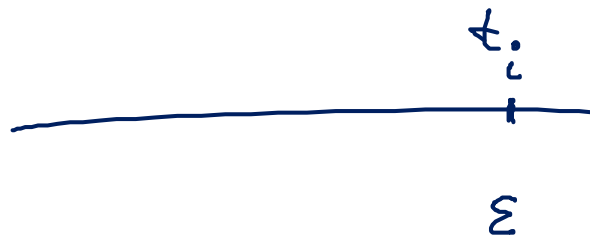
↓ if  $t_b = p_i$

Base cases:

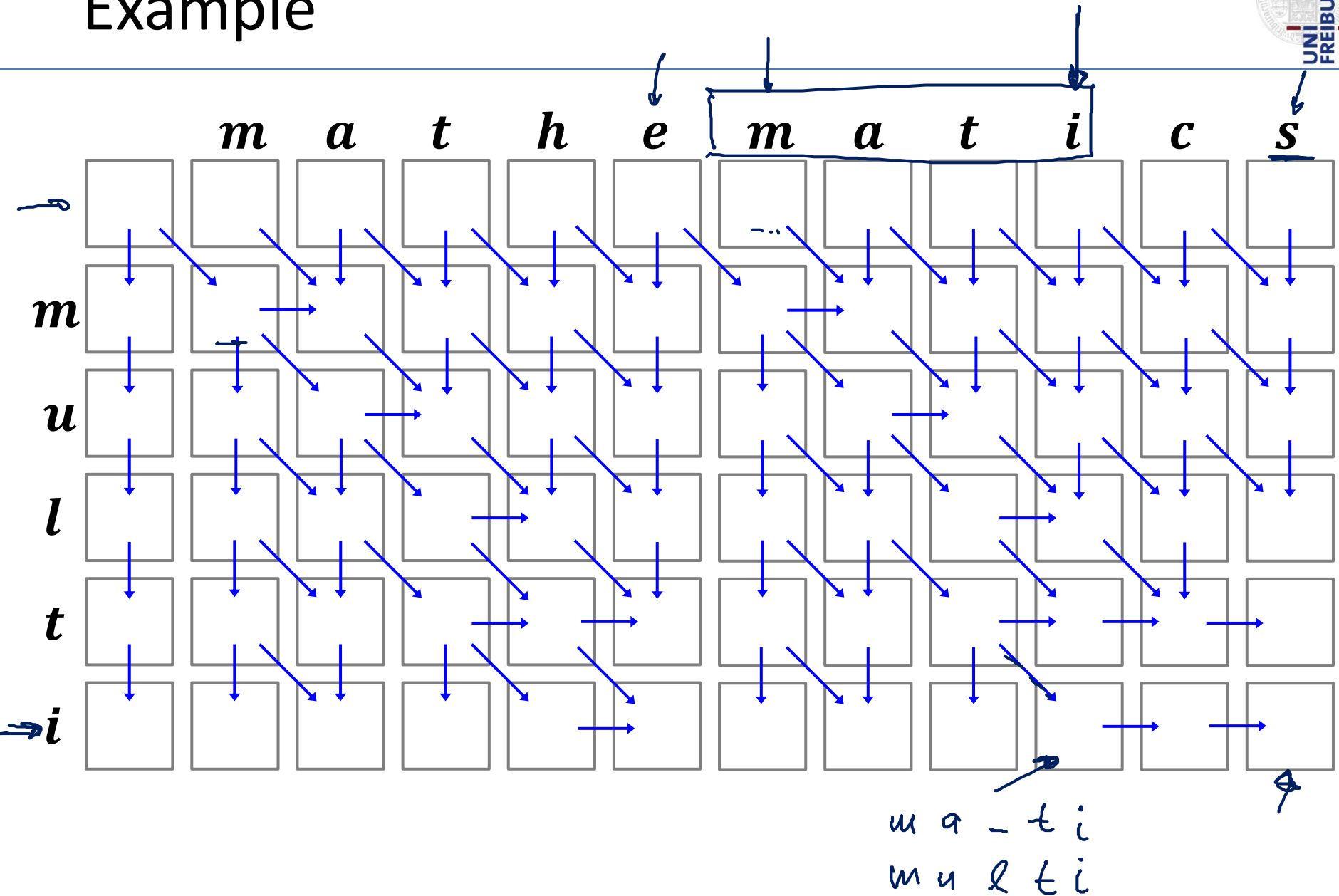
$$\underline{E_{0,0} = 0}$$

$$\underline{E_{0,i} = i}$$

$$\underline{E_{i,0} = 0}$$



# Example





# Approximate String Matching

- Optimal matching consists of optimal sub-matchings
- Optimal matching can be computed in  $O(mn)$  time
- Get matching(s):
  - Start from minimum entry/entries in bottom row
  - Follow path(s) to top row
- Algorithm to compute  $E(b, i)$  identical to edit distance algorithm, except for the initialization of  $E(b, 0)$

## Sequence Alignment:

Find optimal alignment of two given DNA, RNA, or amino acid sequences.

```
G A - C G G A T T A G
G A T C G G A A T - G
```

## Global vs. Local Alignment:

- *Global alignment*: find optimal alignment of 2 sequences
- *Local alignment*: find optimal alignment of sequence 1 (patter) with sub-sequence of sequence 2 (text)