

# TEMPLATE MATCHING

- ❖ The Goal: Given a set of **reference patterns** known as **TEMPLATES**, find to which one an unknown pattern matches best. That is, each class is represented by a **single typical** pattern.
- ❖ The crucial point is to adopt an appropriate "**measure**" to quantify similarity or matching.
- ❖ These measures must accommodate, in an efficient way, deviations between the template and the **test pattern**. For example the word **beauty** may have been read a **beeauty** or **beuty**, etc., due to errors.

P1: 3.5, 0.9, 0.1 G

P2: 2.5 0.6, 0.4 B

P3: 3.8, 1.0, 0.0 G

Reference Patterns|

Unknown Pattern

Pu: 2.9, 0.3, 0.5 ?  $P^I$

Pu vs P1, 0.6, 0.6, 0.4

Pu vs P2 0.4, 0.3, 0.1

Pu vs P3 0.9, 0.7, 0.5

## ❖ Typical Applications

- Speech Recognition
- Motion Estimation in Video Coding
- Data Base Image Retrieval
- Written Word Recognition
- Bioinformatics

## ❖ Measures based on optimal path searching techniques

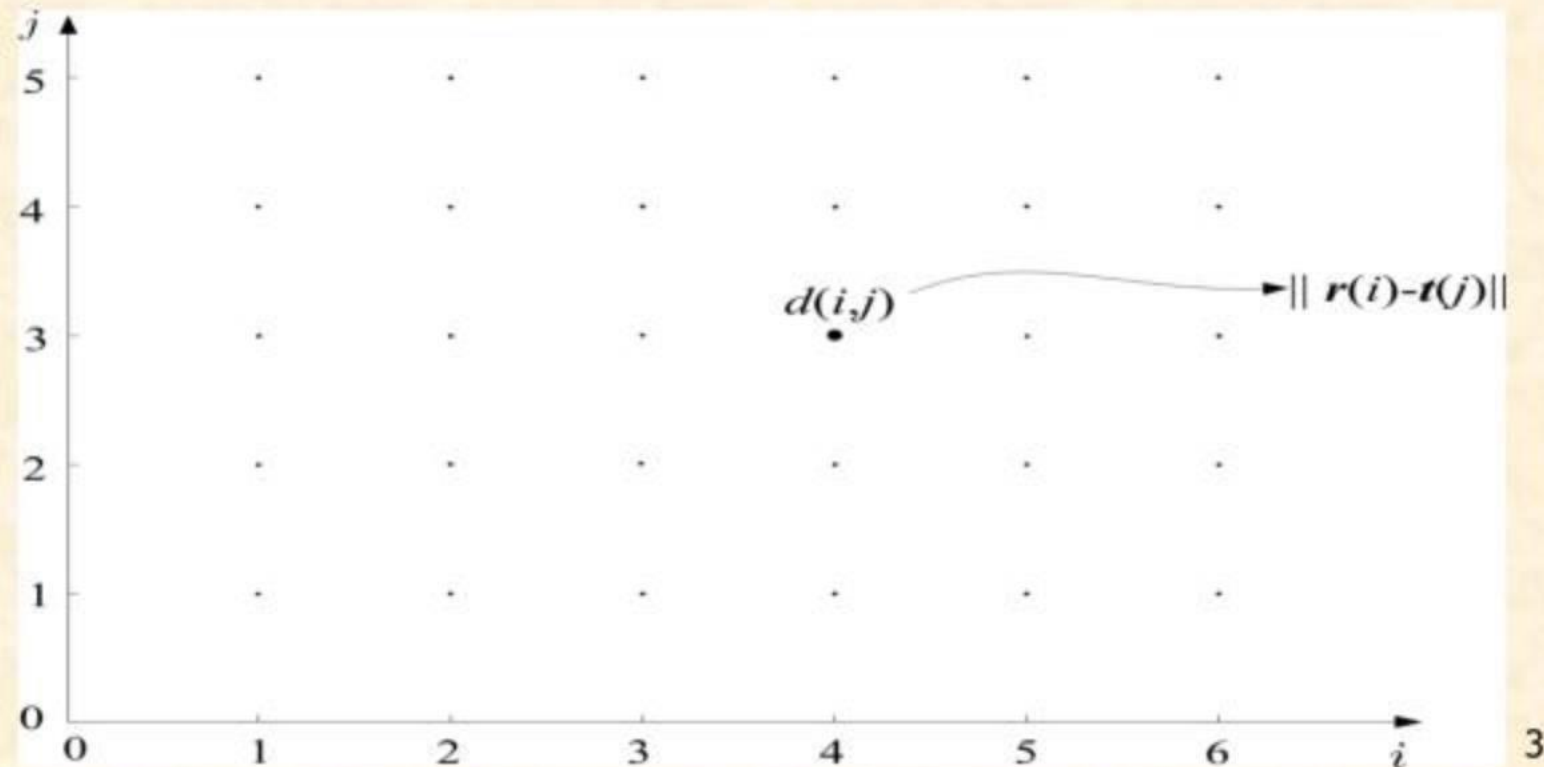
- Representation: Represent the template by a **sequence** of measurement vectors

**Template:**  $\underline{r}(1), \underline{r}(2), \dots, \underline{r}(I)$

**Test pattern:**  $\underline{t}(1), \underline{t}(2), \dots, \underline{t}(J)$



- In general  $I \neq J$
- Form a grid with  $I$  points (template) in horizontal and  $J$  points (test) in vertical
- Each point  $(i,j)$  of the grid measures the distance between  $\underline{r}(i)$  and  $\underline{t}(j)$

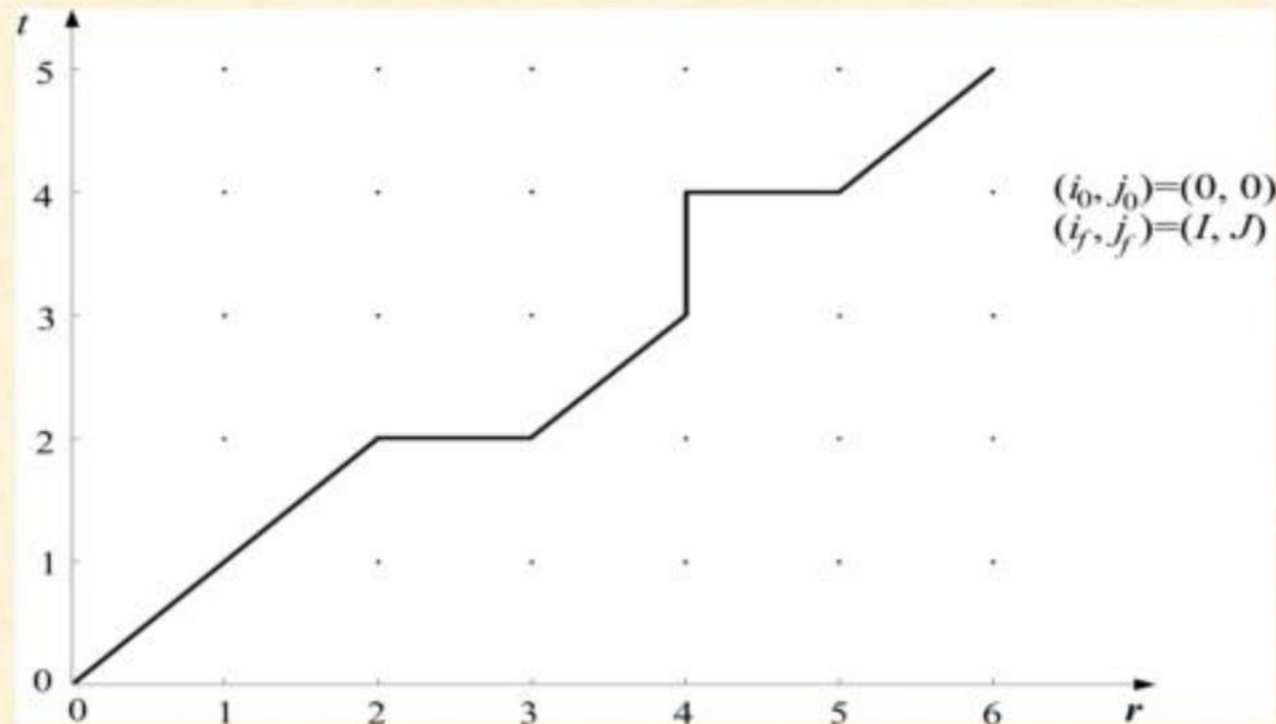


➤ **Path:** A path through the grid, from an initial node  $(i_0, j_0)$  to a final one  $(i_f, j_f)$ , is an **ordered set** of nodes  $(i_0, j_0), (i_1, j_1), (i_2, j_2) \dots (i_k, j_k) \dots (i_f, j_f)$

➤ Each path is associated with a cost

$$D = \sum_{k=0}^{K-1} d(i_k, j_k)$$

where  $K$  is the number of nodes across the path



- Search for the path with the optimal cost  $D_{opt}$ .
- The matching cost between template  $\underline{r}$  and test pattern  $\underline{t}$  is  $D_{opt}$ .

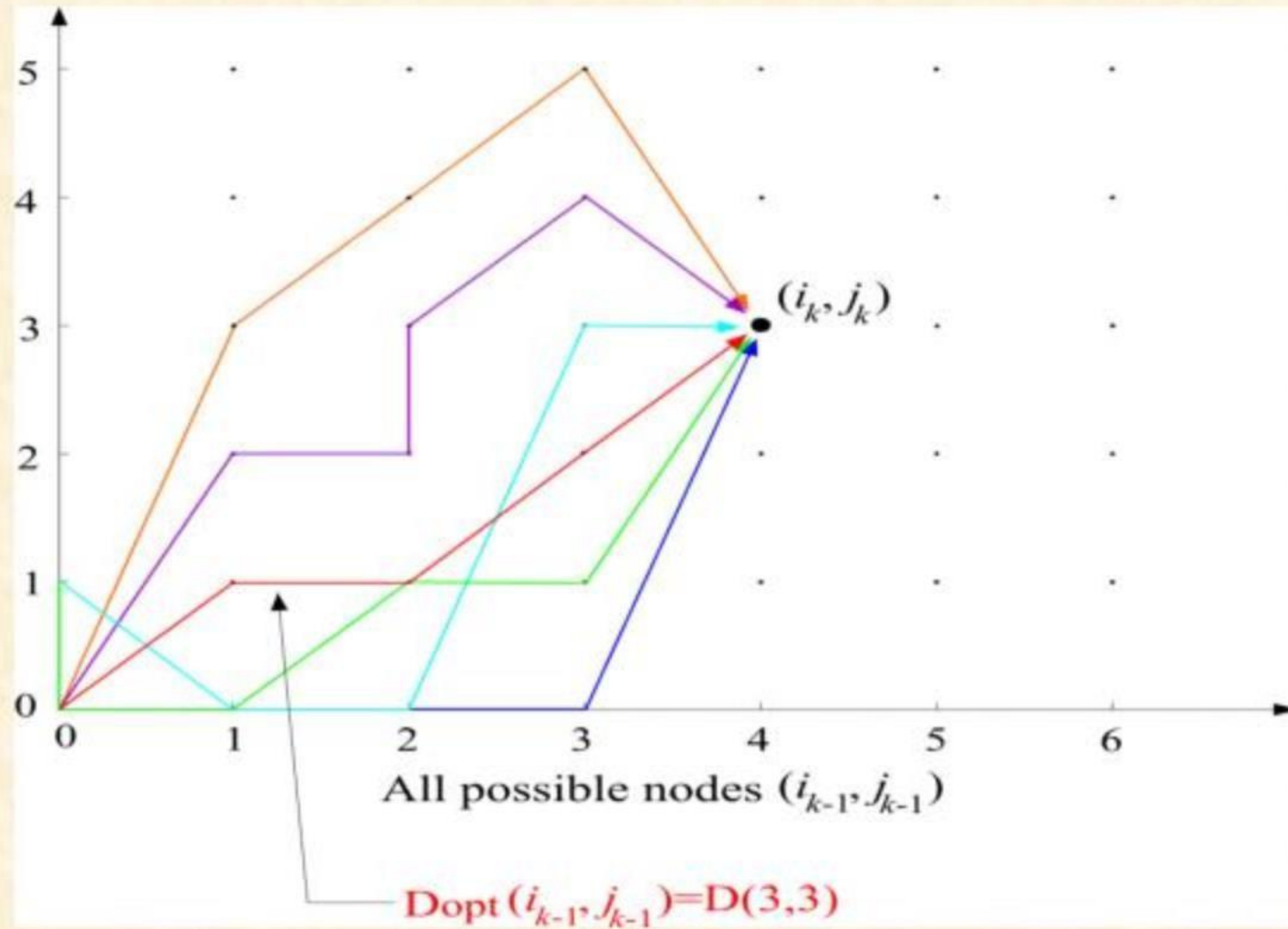
❖ **Bellman's Principle:**

$$(i_0, j_0) \xrightarrow{opt} (i_f, j_f) = (i_0, j_0) \xrightarrow{opt} (i, j) \oplus (i, j) \xrightarrow{opt} (i_f, j_f)$$

- ❖ In words: The **overall** optimal path from  $(i_0, j_0)$  to  $(i_f, j_f)$  **through**  $(i, j)$  is the **concatenation** of the optimal paths from  $(i_0, j_0)$  to  $(i, j)$  **and** from  $(i, j)$  to  $(i_f, j_f)$
- ❖ Let  $D_{opt.}(i, j)$  is the optimal path to reach  $(i, j)$  from  $(i_0, j_0)$ , then Bellman's principle is stated as:



$$D_{opt}(i_k, j_k) = opt\{D_{opt}(i_{k-1}, j_{k-1}) + d(i_k, j_k)\}$$





### ❖ The Edit distance

- It is used for matching written words.

Applications:

- Automatic Editing
- Text Retrieval

- ❖ The cost is based on the philosophy behind the so-called **variational similarity**, i.e.,
  - Measure the cost associated with **converting** one **pattern to the other**
- ❖ Edit distance: **Minimal** total number of **changes**,  **$C$** , **insertions**  **$I$**  and **deletions**  **$R$** , required to change pattern  $A$  into pattern  $B$ ,

$$D(A, B) = \min_j [C(j) + I(j) + R(j)]$$

where  $j$  runs over **All** possible variations of symbols, in order to convert  $A \longrightarrow B$

❖ Allowable predecessors and costs

➤  $(i-1, j-1) \rightarrow (i, j)$

$$d(i, j | i-1, j-1) = \begin{cases} 0, & \text{if } t(i) = r(j) \\ 1, & t(i) \neq r(j) \end{cases}$$

➤ Horizontal (insert)

$$d(i, j | i-1, j) = 1$$

➤ Vertical (Delete)

$$d(i, j | i, j-1) = 1$$





## Question

Using the Edit Distance algorithm, find the minimum edit distance between the words “**intention**” and “**execution**”.

**Show all steps and the final DP table.**

## ❖ The Algorithm

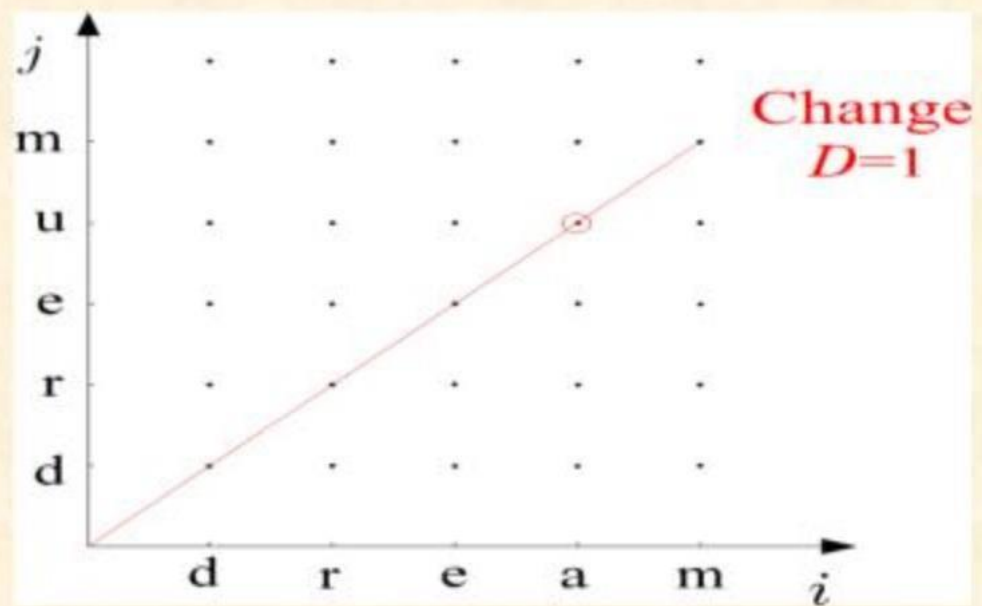
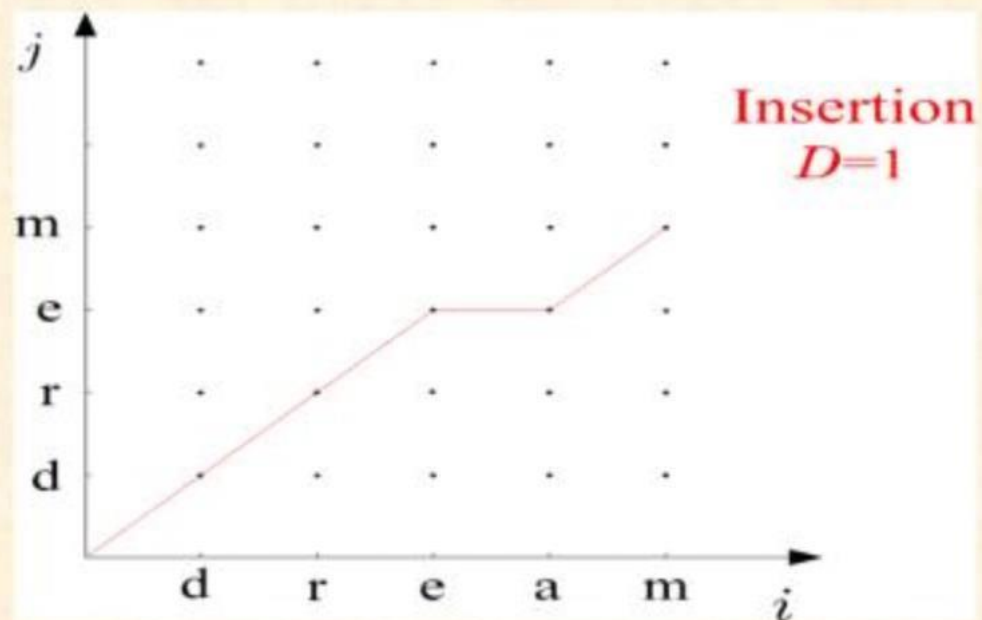
- $D(0,0)=0$
- For  $i=1$ , to  $I$ 
  - $D(i,0)=D(i-1,0)+1$
- END {FOR}
- For  $j=1$  to  $J$ 
  - $D(0,j)=D(0,j-1)+1$
- END{FOR}
- For  $i=1$  to  $I$ 
  - For  $j=1$ , to  $J$ 
    - $C_1=D(i-1,j-1)+d(i,j \mid i-1,j-1)$
    - $C_2=D(i-1,j)+1$
    - $C_3=D(i,j-1)+1$
    - $D(i,j)=\min (C_1,C_2,C_3)$
  - END {FOR}
- END {FOR}
- $D(A,B)=D(I,J)$



## Question

Using the edit distance algorithm, draw the paths for matching the three words '**drem**', '**dreaaam**', '**dreum**' to the actual word '**dream**'.

❖ Examples:



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