

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

P_u: 2.9, 0.3, 0.5 ? P^I

P_u vs P1, 0.6, 0.6, 0.4

P_u vs P2 0.4, 0.3, 0.1

P_u 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

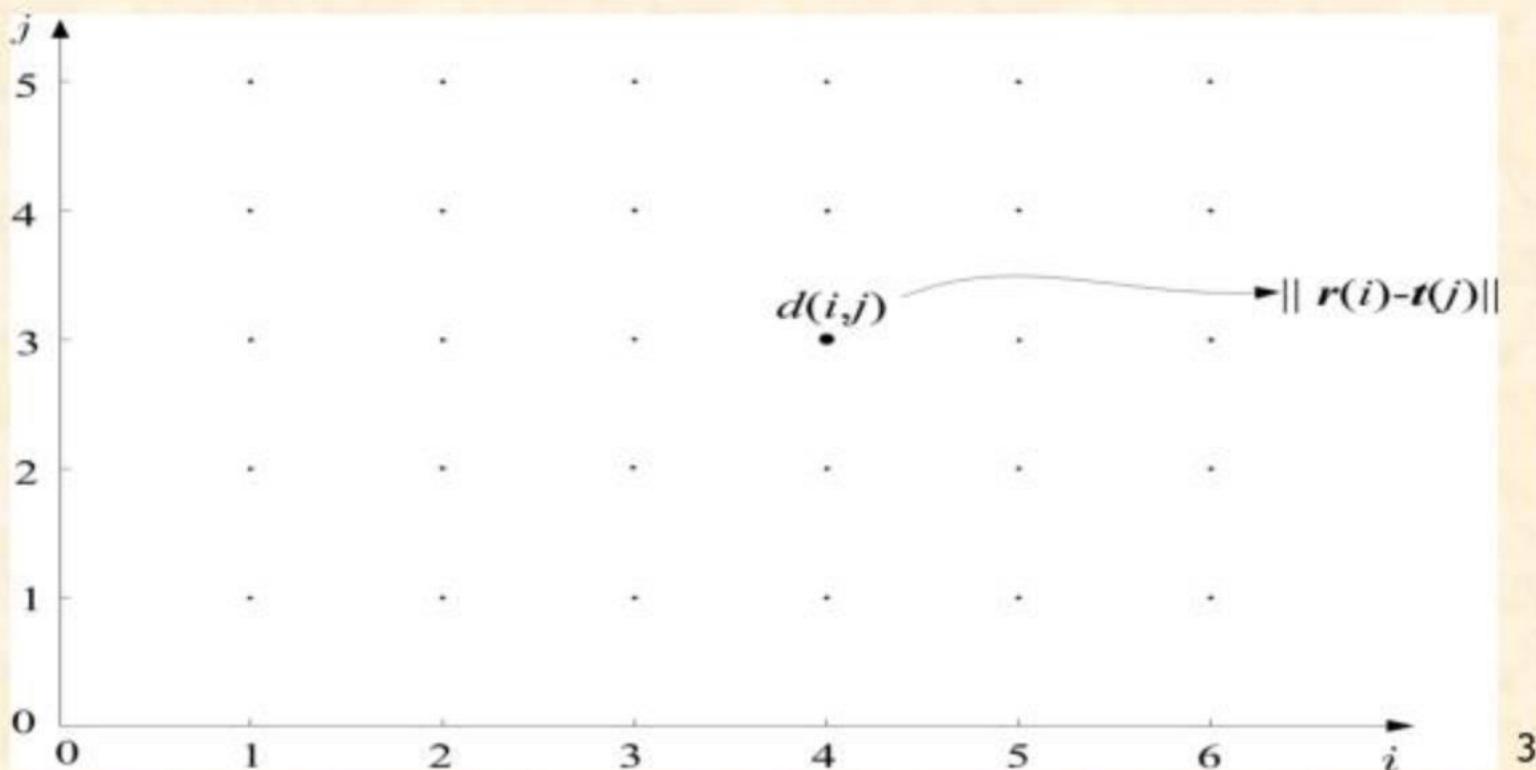
❖ Measures based on optimal path searching techniques

- 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 $\underline{t}(j)$

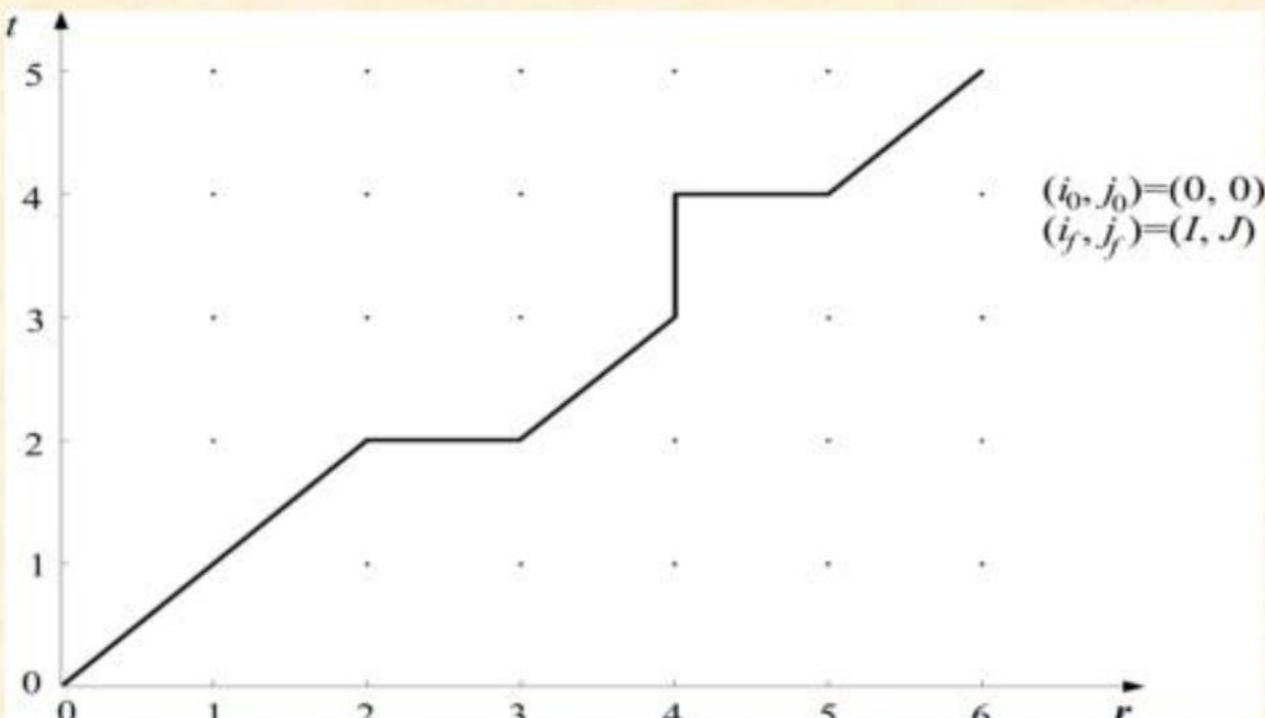


➤ **Path:** A path through the grid, from an initial node (i_0, j_0) to a final one (i_f, j_f) , is an ordered set of nodes $(i_0, j_0), (i_1, j_1), (i_2, j_2) \dots (i_k, j_k) \dots (i_f, j_f)$

➤ 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



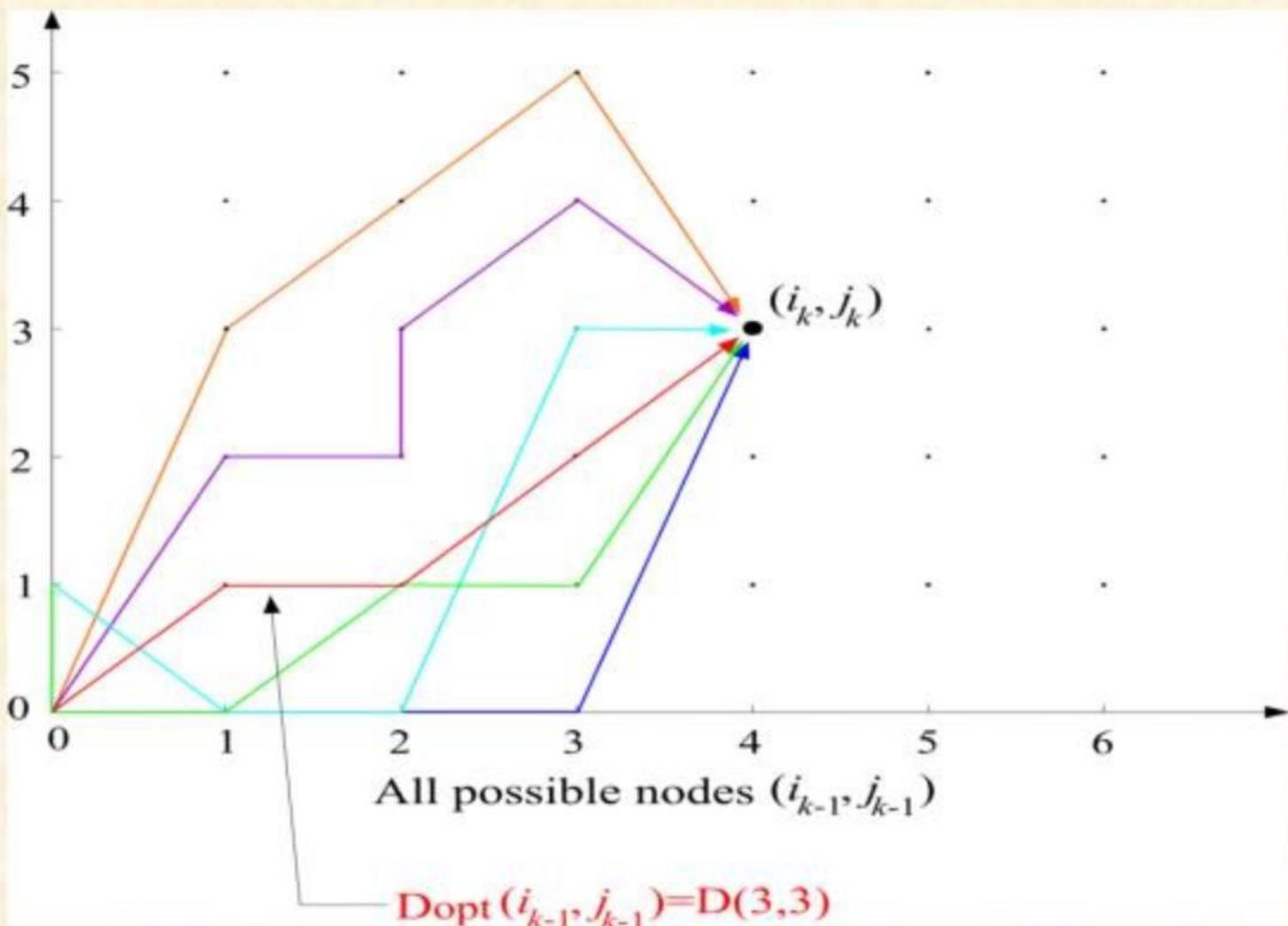
- Search for the path with the optimal cost $D_{opt.}$.
- The matching cost between template \underline{r} and test pattern \underline{t} is $D_{opt.}$

❖ Bellman's Principle:

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

- ❖ 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 $D_{\text{opt.}}(i, j)$ is the optimal path to reach (i, j) from (i_0, j_0) , then Bellman's principle is stated as:

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



❖ 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

➤ $(i-1, j-1) \rightarrow (i, j)$

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

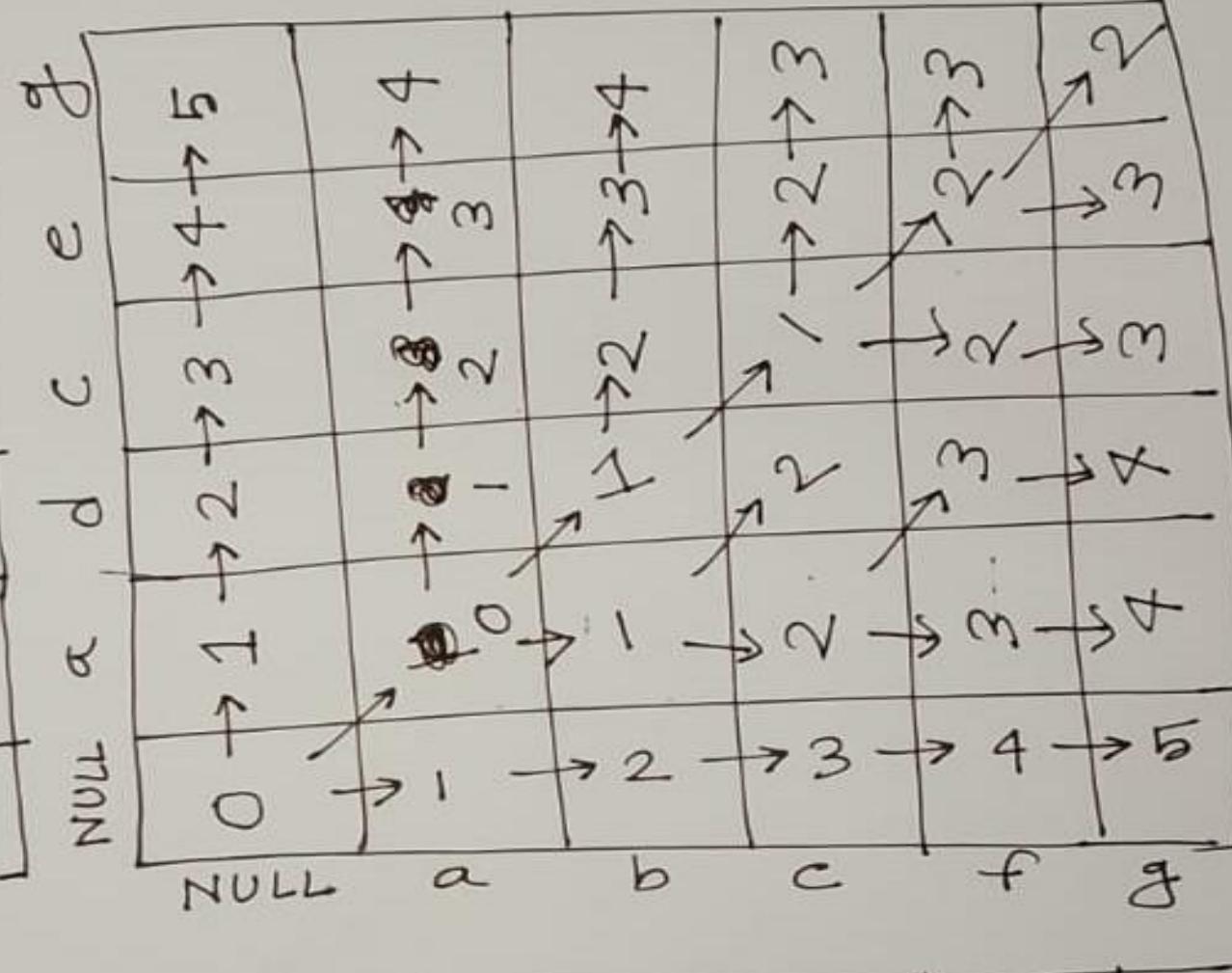
➤ Horizontal (insert)

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

➤ Vertical (Delete)

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

0	-	2	3	4	5
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e.

0	1	2	3	4	5
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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.

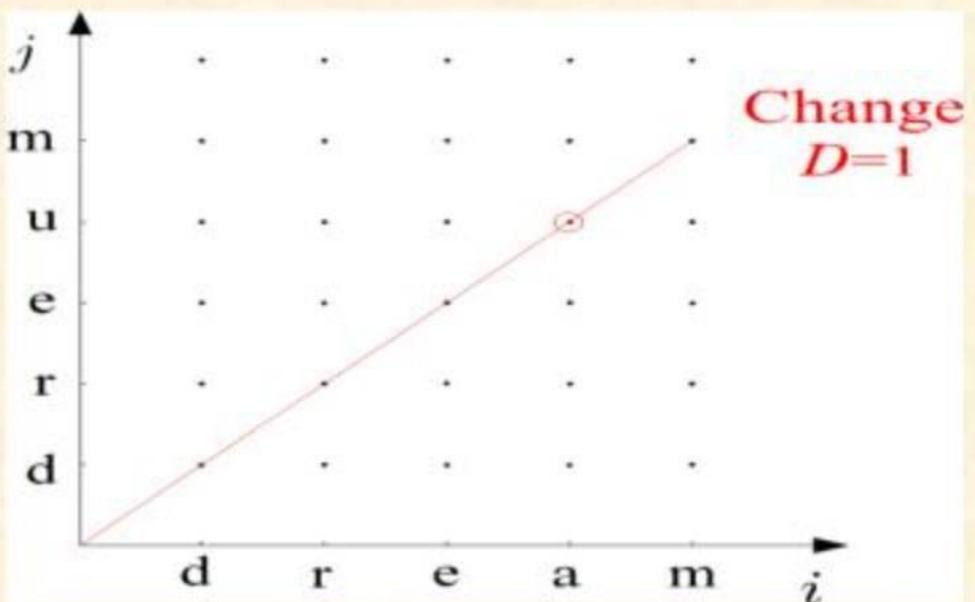
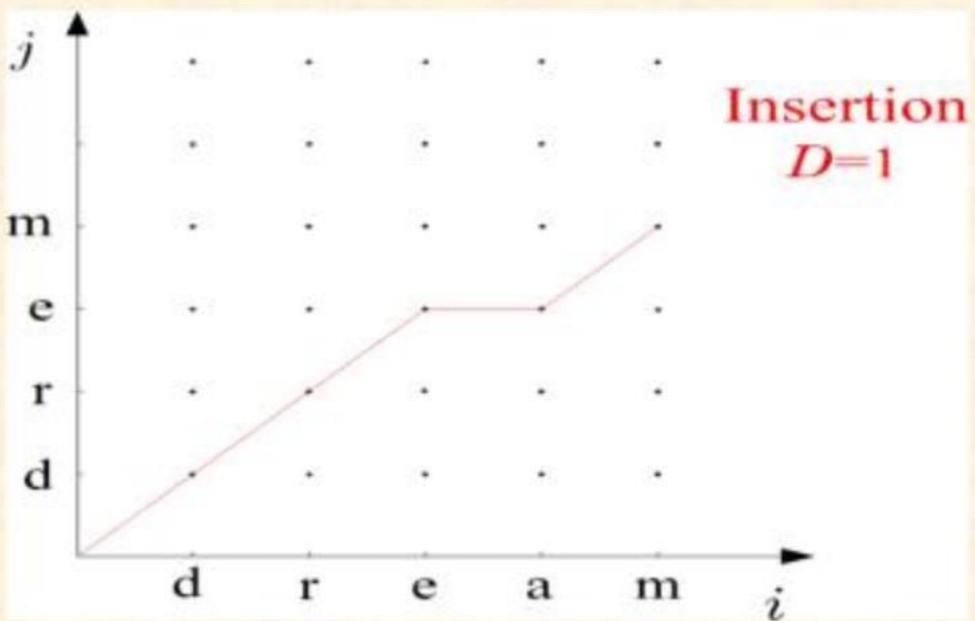
❖ The Algorithm

- $D(0,0)=0$
- 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
 - $C_1=D(i-1,j-1)+d(i,j + i-1,j-1)$
 - $C_2=D(i-1,j)+1$
 - $C_3=D(i,j-1)+1$
 - $D(i,j)=\min(C_1, C_2, C_3)$
- 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**', '**dreaam**', '**dreum**' to the actual word '**dream**'.

❖ Examples:



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