

# Natural Language Processing Laboratory (CS 753)

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# Minimum Edit Distance

## Minimum Edit Distance

- ▶ minimum edit distance between two strings is defined as the minimum number of editing operations needed to transform one into the other.

The editing operations like:

- ▶ insertion
- ▶ deletion
- ▶ substitution

Example: Representing the minimum edit distance between two strings as an alignment.

```

I N T E * N T I O N
| | | | | | | |
* E X E C U T I O N
d s s   i s

```

## The Minimum Edit Distance Algorithm

- ▶ Given two strings, the source string  $X$  of length  $n$ , and target string  $Y$  of length  $m$ , we'll define  $D[i, j]$  as the edit distance between  $X[1 \dots i]$  and  $Y[1 \dots j]$ , i.e., the first  $i$  characters of  $X$  and the first  $j$  characters of  $Y$ . The edit distance between  $X$  and  $Y$  is thus  $D[n, m]$ .

## The Minimum Edit Distance Algorithm

- ▶ **Dynamic programming:** A tabular computation of  $D(n, m)$
- ▶ Solving problems by combining solutions to subproblems.
- ▶ Bottom-up
  - ▶ We compute  $D(i, j)$  for small  $i, j$
  - ▶ And compute larger  $D(i, j)$  based on previously computed smaller values
  - ▶ i.e., compute  $D(i, j)$  for all  $i (0 < i < n)$  and  $j (0 < j < m)$ .

## The Minimum Edit Distance Algorithm

- ▶ use dynamic programming to compute  $D[n, m]$  bottom up, combining solutions to subproblems.

$$D[i, j] = \min \begin{cases} D[i-1, j] + \text{del-cost}(\text{source}[i]) \\ D[i, j-1] + \text{ins-cost}(\text{target}[j]) \\ D[i-1, j-1] + \text{sub-cost}(\text{source}[i], \text{target}[j]) \end{cases}$$

- ▶ assume the version of Levenshtein distance in which the insertions and deletions each have a cost of 1 and substitutions have a cost of 2

$$D[i, j] = \min \begin{cases} D[i-1, j] + 1 \\ D[i, j-1] + 1 \\ D[i-1, j-1] + \begin{cases} 2; & \text{if } \text{source}[i] \neq \text{target}[j] \\ 0; & \text{if } \text{source}[i] = \text{target}[j] \end{cases} \end{cases}$$

# The Minimum Edit Distance Algorithm

**function** MIN-EDIT-DISTANCE(*source*, *target*) **returns** *min-distance*

$n \leftarrow \text{LENGTH}(\text{source})$

$m \leftarrow \text{LENGTH}(\text{target})$

Create a distance matrix  $\text{distance}[n+1, m+1]$

# Initialization: the zeroth row and column is the distance from the empty string

$D[0,0] = 0$

**for** each row  $i$  **from** 1 **to**  $n$  **do**

$D[i,0] \leftarrow D[i-1,0] + \text{del-cost}(\text{source}[i])$

**for** each column  $j$  **from** 1 **to**  $m$  **do**

$D[0,j] \leftarrow D[0,j-1] + \text{ins-cost}(\text{target}[j])$

# Recurrence relation:

**for** each row  $i$  **from** 1 **to**  $n$  **do**

**for** each column  $j$  **from** 1 **to**  $m$  **do**

$D[i,j] \leftarrow \text{MIN}( D[i-1,j] + \text{del-cost}(\text{source}[i]),$   
 $D[i-1,j-1] + \text{sub-cost}(\text{source}[i], \text{target}[j]),$   
 $D[i,j-1] + \text{ins-cost}(\text{target}[j]))$

# Termination

**return**  $D[n,m]$

## The Minimum Edit Distance(MED) Algorithm

- Computation of MED between *intention* and *execution*:

Src\Tar	#	e	x	e	c	u	t	i	o	n
#	0	1	2	3	4	5	6	7	8	9
i	1	2	3	4	5	6	7	6	7	8
n	2	3	4	5	6	7	8	7	8	7
t	3	4	5	6	7	8	7	8	9	8
e	4	3	4	5	6	7	8	9	10	9
n	5	4	5	6	7	8	9	10	11	10
t	6	5	6	7	8	9	8	9	10	11
i	7	6	7	8	9	10	9	8	9	10
o	8	7	8	9	10	11	10	9	8	9
n	9	8	9	10	11	12	11	10	9	8



## Computing alignments

- ▶ Edit distance isn't sufficient
  - ▶ We often need to **align** each character of the two strings to each other
- ▶ We do this by keeping a “backtrace”
- ▶ Every time we enter a cell, remember where we came from
- ▶ When we reach the end
  - ▶ Trace back the path from the **lower right** corner to read off the alignment

## Alignments and edit distance

These two problems reduce to one: find the **optimal character alignment** between two words (the one with the fewest character changes: the **minimum edit distance** or **MED**).

► Example:

if all changes count equally, **MED(stall, table)** is 3

S	T	A	L	L	-	
	T	A	L	L		deletion
	T	A	B	L		substitution
	T	A	B	L	E	insertion

► Written as an alignment:

S	T	A	L	L	-
d			s		i
	T	A	B	L	E

## More Alignments

- There may be multiple best alignments. In this case, two:

S	T	A	L	L	-	S	T	A	-	L	L
d			s		i	d			i		s
	T	A	B	L	E		T	A	B	L	E

- And lots of non-optimal alignments, such as:

S	T	A	-	L	L
s	s	s	i	s	d
T	A	B	L	E	

## Assignments:

1. Given two strings, the source string  $X$  of length  $n$ , and target string  $Y$  of length  $m$ , define  $D[i, j]$  as the edit distance between  $X[1 \dots i]$  and  $Y[1 \dots j]$ , i.e., the first  $i$  characters of  $X$  and the first  $j$  characters of  $Y$ . The edit distance between  $X$  and  $Y$  is thus  $D[n, m]$ . Write a program to compute the edit distance,  $D[n, m]$  between  $X$  and  $Y$  using the MED algorithm as discussed in the class.
2. Choose a different path through the backpointers and reconstruct its alignment. How many different optimal alignments are there? Show all. Use your hand-computed results to check your code.

## References

- ▶ Steven Bird, Ewan Klein, and Edward Loper, “Natural Language Processing with Python”, Published by O'Reilly Media, Inc.