

Minimum Edit Distance

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How similar are two strings?

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- Spell correction

- The user typed “graffe”

Which is closest?

- ☐ graf
 - ☐ graft
 - ☐ grail
 - ☐ giraffe

- Also for Machine Translation, Information Extraction, Speech Recognition

Edit Distance

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- The minimum edit distance between two strings
- Is the minimum number of editing operations
 - Insertion
 - Deletion
 - Substitution
- Needed to transform one into the other

Minimum Edit Distance

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- Two strings and their **alignment**:

I	N	T	E	*	N	T	I	O	N
*	E	X	E	C	U	T	I	O	N

Minimum Edit Distance

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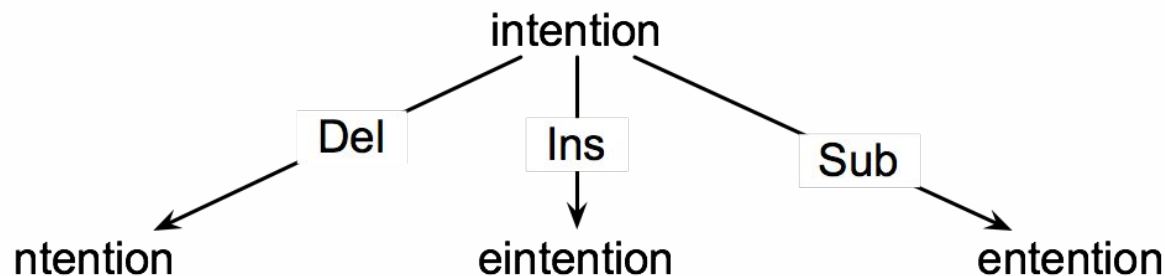
I N T E * N T I O N
| | | | | | | | | |
* E X E C U T I O N
d s s i s

- If each operation has cost of 1
 - Distance between these is 5
- If substitutions cost 2 (Levenshtein)
 - Distance between them is 8

How to find the Min Edit Distance?

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- Searching for a path (sequence of edits) from the start string to the final string:
 - **Initial state:** the word we're transforming
 - **Operators:** insert, delete, substitute
 - **Goal state:** the word we're trying to get to
 - **Path cost:** what we want to minimize: the number of edits



Minimum Edit as Search

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- But the space of all edit sequences is huge!
 - We can't afford to navigate naïvely
 - Lots of distinct paths wind up at the same state.
 - We don't have to keep track of all of them
 - Just the shortest path to each of those revisited states.

Defining Min Edit Distance

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- For two strings
 - X of length n
 - Y of length m
- We define $D(i,j)$
 - the edit distance between $X[1..i]$ and $Y[1..j]$
 - i.e., the first i characters of X and the first j characters of Y
 - The edit distance between X and Y is thus $D(n,m)$

Dynamic Programming for Minimum Edit Distance

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- **Dynamic programming:** A tabular computation of $D(n,m)$
- Solving problems by combining solutions to sub problems.
- 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 ($0 < j < m$)

Defining Min Edit Distance (Levenshtein)

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- Initialization

$$D(i, 0) = i$$

$$D(0, j) = j$$

- Recurrence Relation:

For each $i = 1 \dots M$

For each $j = 1 \dots N$

$$D(i, j) = \begin{cases} \min \begin{cases} D(i-1, j) + 1 \\ D(i, j-1) + 1 \\ D(i-1, j-1) + 2 \end{cases} & \text{if } X(i) \neq Y(j) \\ 0 & \text{if } X(i) = Y(j) \end{cases}$$

- Termination:

$D(N, M)$ is distance

The Edit Distance Table: Initialization

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N	9									
O	8									
I	7									
T	6									
N	5									
E	4									
T	3									
N	2									
I	1									
#	0	1	2	3	4	5	6	7	8	9
	#	E	X	E	C	U	T	I	O	N

The Edit Distance Table



N	9									
O	8									
I	7									
T	6									
N	5									
E	4									
T	3									
N	2									
I	1	2								
#	0	1	2	3	4	5	6	7	8	9
	#	E	X	E	C	U	T	I	O	N

$$D(i,j) = \min \begin{cases} D(i-1,j) + 1 \\ D(i,j-1) + 1 \\ D(i-1,j-1) + \begin{cases} 2; & \text{if } S_1(i) \neq S_2(j) \\ 0; & \text{if } S_1(i) = S_2(j) \end{cases} \end{cases}$$



The Edit Distance Table



N	9	8	9	10	11	12	11	10	9	8
O	8	7	8	9	10	11	10	9	8	9
I	7	6	7	8	9	10	9	8	9	10
T	6	5	6	7	8	9	8	9	10	11
N	5	4	5	6	7	8	9	10	11	10
E	4	3	4	5	6	7	8	9	10	9
T	3	4	5	6	7	8	7	8	9	8
N	2	3	4	5	6	7	8	7	8	7
I	1	2	3	4	5	6	7	6	7	8
#	0	1	2	3	4	5	6	7	8	9
	#	E	X	E	C	U	T	I	O	N

Computing alignments

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- 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 upper right corner to read off the alignment

Edit Distance

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N	9									
O	8									
I	7									
T	6									
N	5									
E	4									
T	3									
N	2									
I	1									
#	0	1	2	3	4	5	6	7	8	9
	#	E	X	E	C	U	T	I	O	N

$$D(i,j) = \min \begin{cases} D(i-1,j) + 1 \\ D(i,j-1) + 1 \\ D(i-1,j-1) + \begin{cases} 2; & \text{if } S_1(i) \neq S_2(j) \\ 0; & \text{if } S_1(i) = S_2(j) \end{cases} \end{cases}$$

MinEdit with Backtrace

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

n	9	↓ 8	↙↖↓ 9	↙↖↓ 10	↙↖↓ 11	↙↖↓ 12	↓ 11	↓ 10	↓ 9	↙ 8	
o	8	↓ 7	↙↖↓ 8	↙↖↓ 9	↙↖↓ 10	↙↖↓ 11	↓ 10	↓ 9	↙ 8	← 9	
i	7	↓ 6	↙↖↓ 7	↙↖↓ 8	↙↖↓ 9	↙↖↓ 10	↓ 9	↙ 8	← 9	← 10	
t	6	↓ 5	↙↖↓ 6	↙↖↓ 7	↙↖↓ 8	↙↖↓ 9	↙ 8	← 9	← 10	↖↓ 11	
n	5	↓ 4	↙↖↓ 5	↙↖↓ 6	↙↖↓ 7	↙↖↓ 8	↙↖↓ 9	↙↖↓ 10	↙↖↓ 11	↙↓ 10	
e	4	↙ 3	← 4	↙↖ 5	← 6	← 7	↖↓ 8	↙↖↓ 9	↙↖↓ 10	↓ 9	
t	3	↙↖↓ 4	↙↖↓ 5	↙↖↓ 6	↙↖↓ 7	↙↖↓ 8	↙ 7	↖↓ 8	↙↖↓ 9	↓ 8	
n	2	↙↖↓ 3	↙↖↓ 4	↙↖↓ 5	↙↖↓ 6	↙↖↓ 7	↙↖↓ 8	↓ 7	↙↖↓ 8	↙ 7	
i	1	↙↖↓ 2	↙↖↓ 3	↙↖↓ 4	↙↖↓ 5	↙↖↓ 6	↙↖↓ 7	↙ 6	← 7	← 8	
#	0	1	2	3	4	5	6	7	8	9	
	#	e	x	e	c	u	t	i	o	n	

Adding Backtrace to Minimum Edit Distance

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- Base conditions:

$$D(i, 0) = i$$

$$D(0, j) = j$$

$D(N, M)$ is distance

- Recurrence Relation:

For each $i = 1 \dots M$

For each $j = 1 \dots N$

$$D(i, j) = \min$$

2; if $X(i) \neq Y(j)$

0; if $X(i) = Y(j)$

$ptr(i, j) =$

insertion

$$D(i-1, j) + 1$$

deletion

$$D(i, j-1) + 1$$

$$D(i-1, j-1) +$$

insertion

substitution

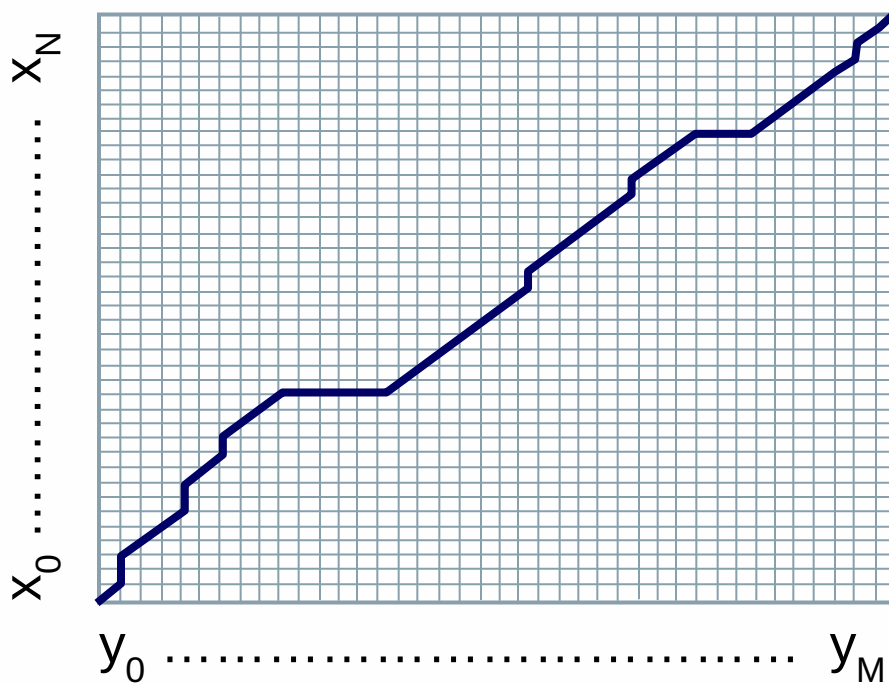
deletion

substitution

LEFT
DOWN
DIAG

The Distance Matrix

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Every non-decreasing path

from $(0,0)$ to (M, N)

corresponds to
an alignment
of the two sequences

An optimal alignment is composed
of optimal sub alignments

Result of Backtrace

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- Two strings and their **alignment**:

I	N	T	E	*	N	T	I	O	N
*	E	X	E	C	U	T	I	O	N

Performance

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- Time:

$O(nm)$

- Space:

$O(nm)$

- Backtrace

$O(n+m)$

THANK YOU