



EDIT DISTANCE

THEORY & APPLICATIONS OF NLP
WOA7013






SPELLING ERRORS DETECTION & CORRECTION



non-word error detection: detecting spelling errors which result in non-words (e.g. *tomrrow* for *tomorrow*).

isolated-word error correction: correcting spelling errors which result in non-words, e.g. correcting *tomrrow* to *tomorrow*, but looking only at the word in isolation.



context-dependent error detection and correction: Using the context to help detect and correct spelling errors even if they accidentally result in an actual word (real-word errors). This REAL-WORD ERRORS can happen from typographical errors (insertion, deletion, transposition) which accidentally produce a real word (e.g. *there* for *three*), or because the writer substituted the wrong spelling of a homophones (e.g. *piece* for *peace*).



INTRODUCTION

How similar are 2 strings?

- Spell Checker/Correction
 - The user typed “markt”. Which is closest?
 - *mark*
 - *make*
 - *market*

- Computational Biology
 - Align 2 sequences of nucleotides

AGGCTATCACCTGACC

TAGCTATCAGCACCGC

- Resulting alignment

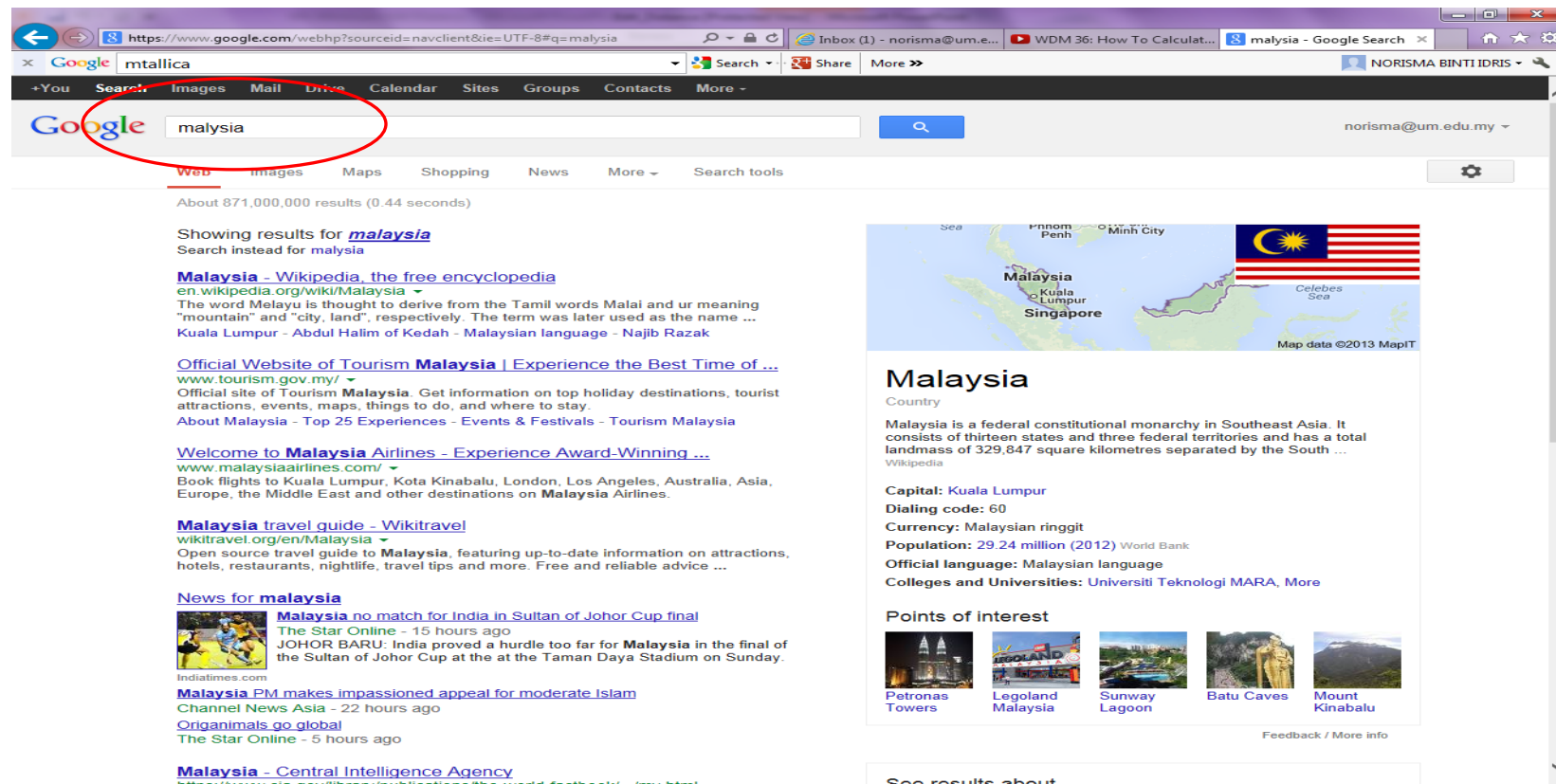
-AGGCTATCACCTGACC--

TAG-CTATCAC--GACCGC



REAL APPLICATION

Searching keywords via the net: usually by “malysia”, we mean “malaysia”





MINIMUM EDIT DISTANCE




Edit distance, also known as **Levenshtein distance** or **Evolutionary distance**.

The min edit distance between 2 strings is the min number of editing operations needed to transform one string into another.

◦ *Given 2 strings, s_1 and s_2 , the minimum number of operations/edits to change from one string to another.*

Operations/edits are typically character-level:

- Deletion (d)
- Insertion (i)
- Substitution (s)

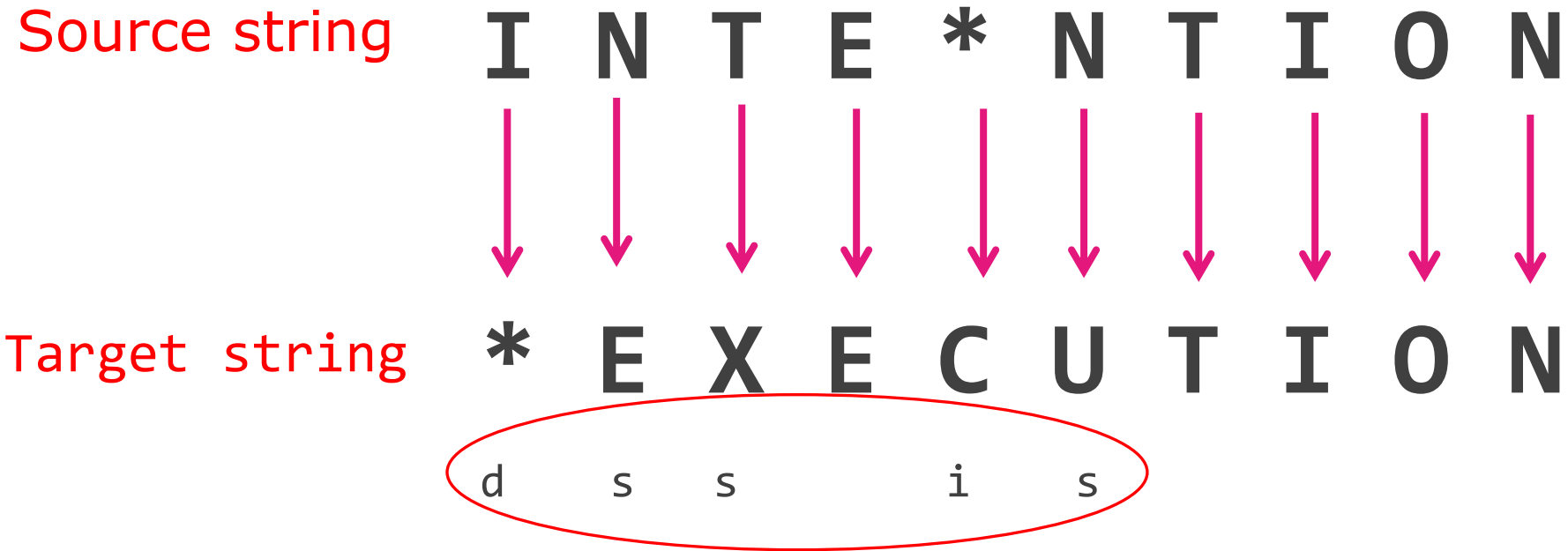


It is the most common measure to expose the dissimilarity between two strings (Levenshtein 1966; Navarro & Raffinot 2002).



For example:

- To find the distance btw 2 words:



The operations
involved



The transformation list from intention to execution:

intention

intention - delete i


intention - substitute n by e

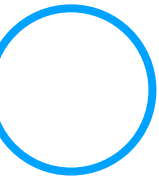
execution - substitute t by x

execution - insert u

execution - substitute n by c


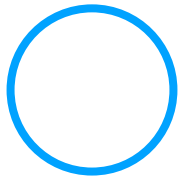
We can assign a particular cost/weight to each of these operations.

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





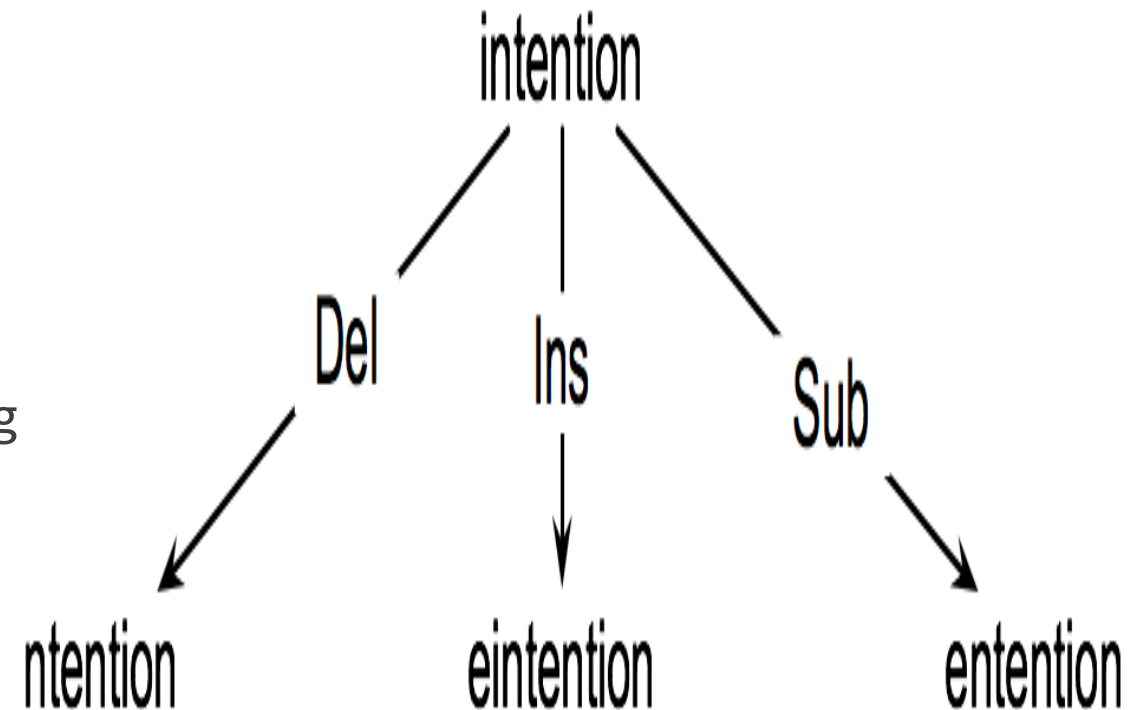
Determine the minimum edit distance between these 2 strings:

1. don and dog = ?
 2. cat and act = ?
 3. cat and dog = ?
- 
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HOW TO FIND A MIN EDIT DISTANCE

Searching for a path (sequence of edit) from the start string to the final string:


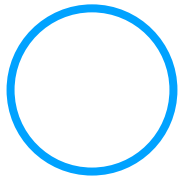
- **Initial state:** the word we're transforming
- **Operators:** *D*, *I*, *S*
- **Goal state:** the word we're trying to get to
- **Path cost:** what we want to minimize: the number of edits





MIN EDIT AS SEARCH

The space of all edit sequences is huge.

- Lots of possible sequences wind up at the same state
 - We don't have to keep track all of them
 - Just the shortest path to each of those revisited states.
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- 


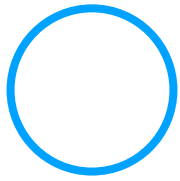


DEFINING MIN EDIT DISTANCE

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 $D(n, m)$
- 
- 



The base cases for edit distance function:



1. $D(0,0) = 0$

$s_1 = 0$ character and
 $s_2 = 0$. Thus, no
operations involved

2. $D(1,0) = 1$

$s_1 = 1$ character and
 $s_2 = 0$. Thus, only 1
operation involved

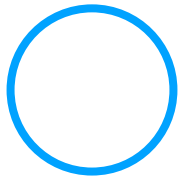
3. $D(i,0) = i$

$s_1 = i$ character and
 $s_2 = 0$. Thus, i
operations involved



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

$s_1 = 0$ character and
 $s_2 = j$. Thus, j
operations involved




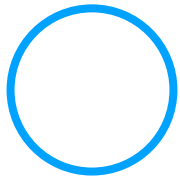


COMPUTING MIN EDIT DISTANCE

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 ($0 < j < m$)
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MIN EDIT DISTANCE ALGORITHM

Initialization

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

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

Recurrence Relation

For each $i = 1 \dots m$

For each $j = 1 \dots n$

$$D(i, j) = \min \left\{ \begin{array}{ll} D(i-1, j) + 1 & \text{deletion} \\ D(i, j-1) + 1 & \text{insertion} \\ D(i-1, j-1) + 2; & \left\{ \begin{array}{ll} \text{if } X(i) \neq Y(j) & \text{substitution} \\ \emptyset; & \text{if } X(i) = Y(j) \end{array} \right. \end{array} \right.$$

Termination

$D(m, n)$ is the distance


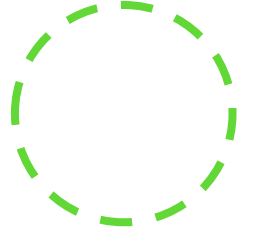
THE EDIT DISTANCE TABLE

N	9									
O	8									
I	7									
T	6									
N	5									
E	4									
T	3									
N	2									
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


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

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



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



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