

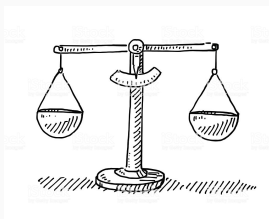
## Edit Distance

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## Definition

### How similar are two strings?



distance between two strings  
↔ how similar are dad and daddy?  
↔ is tomorrow closer to tomorrow or to tomorow?

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### Edit distance

#### String transformation

- given an alphabet  $\Sigma$  and a word  $w \in \Sigma^*$
- Editing operations
  - ↔ insert the character  $x$  at position  $i$
  - ↔ delete the character at position  $i$
  - ↔ substitute the character at position  $i$  by  $x$
- a **sequence** of operations will **transform** a word  $w$  into  $w'$

#### Edit distance

- given two strings  $u, v$ , consider the set of all sequence of editions
- define a cost for each operation (usually 1)
- $\text{distance}(u, v)$  = minimum number of operations to transform  $u$  into  $v$

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### Example

delete i →	i n t e n t i o n
substitute n by e →	n t e n t i o n
substitute t by x →	e t e n t i o n
insert u →	e x e n t i o n
substitute n by c →	e x e n u t i o n
	e x e c u t i o n

- ↔ no shorter sequence of operations
- ↔ edit distance between intention and execution is 5
- ↔ Levenshtein distance ( $\omega_{\text{subst}} = 2$ ): 8

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### Alignment

- the sequence of edit operations define an alignment between the two strings

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

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## Is the edit distance a distance?

If:

- every edit operation has positive cost;
- for every operation, there is an inverse operation with equal cost.

The metric axioms are satisfied as follows:

- $d(a, b) = 0$  if and only if  $a = b$ , each string is trivially transformed to itself with a cost of 0
- $d(a, b) > 0$  when  $a \neq b$ , at least one edit operation
- $d(a, b) = d(b, a)$  by equality of the cost of each operation and its inverse.
- Triangle inequality:  $d(a, c) \leq d(a, b) + d(b, c)$

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## Application

## Comparing documents



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## Identifying Frequent Spelling Correction in French

[Wisniewski et al, TALN'2010]

- compute alignment between the different revision of a wikipedia page
- heuristics to filter modifications that correspond to spelling errors
- compute statistics

Erreurs orthographiques				Erreurs grammaticales			
e → é	6,7%	-l	1,9%	+s	16,2%	-t	1,5%
E → Ê	6,7%	+i	1,9%	+e	9,9%	e → a	1,4%
oe → ø	4,6%	a → ä	1,8%	-s	8,8%	er → é	1,0%
+n	4,3%	-e	1,7%	A → à	5,6%	er → é	0,9%
+s	2,8%	-n	1,7%	-e	4,9%	u → à	0,9%
+r	2,7%	+t	1,6%	i → f	2,7%	à → a	0,9%
é → è	2,7%	+s	1,6%	a → à	2,2%	e → é	0,8%
-s	2,5%	e → à	1,4%	+nt	1,9%	é → à	0,7%
+p	2,2%	+l	1,3%	+t	1,7%	s → t	0,7%
é → e	2,1%	-r	1,3%	a → e	1,5%	à → u	0,7%

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## Identifying Frequent Spelling Correction in French

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- compute alignment between the different revision of a wikipedia page
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	erreurs lexicales	erreurs grammaticales
première moitié du mot	34,06%	4,06%
seconde moitié du mot	62,81%	93,26%
erreurs dans les deux moitiés	3,13%	2,63%

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## Computing minimal edit distance

## Number of sequence of edits

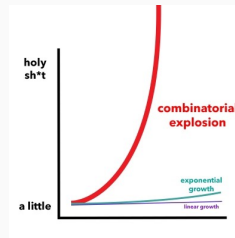
Given two strings  $a$  and  $b$ , how many sequence of edits can transform  $a$  into  $b$ ?

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## Number of sequence of edits

### Observation

- combinatorial explosion
- number of answers grows very/too quickly with respect to the size of the words

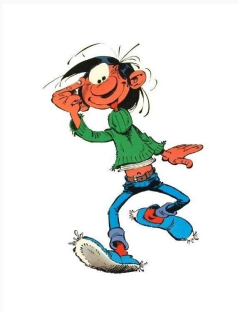


### Consequence

- minimum edit is a **search problem** can be reduced to shortest path problem
- the space of all edit sequences is huge!
- we can afford to navigate

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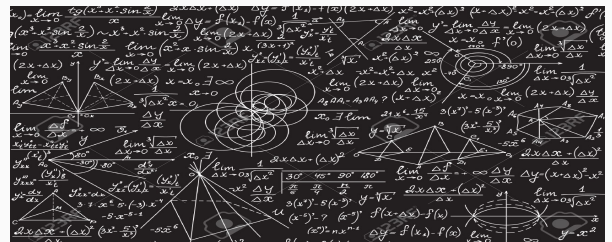
## Some observations



- lots of different sequence of editions result in the same word:
  - $a$  can be transformed into  $b$  by:
    - 1 substitution
    - 1 deletion and 1 insertion
    - ....
 ⇒ overlapping sub-problems
- minimal edit distance → only need to keep track of the 'shortest' way to generate a given word
  - ⇒ optimal substructure

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## Some notations



- $X, Y$ : two strings of length  $n$  and  $m$
- $D(i, j)$ : min. edit distance between  $X[1..i]$  and  $Y[1..j]$ 
  - the first  $i$  characters of  $X$  and  $j$  characters of  $Y$
  - the 'overlapping sub-problems'
- $D(n, m)$  is the edit distance between  $X$  and  $Y$

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## Dynamic Programming



- dynamic programming**: a tabular computation of  $D(n, m)$
- solving problems by combining solutions to sub-problems
- bottom-up:
  - compute  $D(i, j)$  for 'small'  $i, j$
  - compute 'larger'  $D(i, j)$  based on previously computed smaller values
- avoid computing the same 'sub-result' twice

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## An historical wink (1)



- dynamic programming (Bellman, 1950) = generic optimization method
- many applications (e.g. shortest path problem, Viterbi algorithm, matrix multiplication, ...)
- explore an exponential number of solutions in a polynomial time
  - simply by 'organizing' the computation correctly

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## An historical wink (2)

"Where did the name, dynamic programming, come from? The 1950s were not good years for mathematical research. We had a very interesting gentleman in Washington named Wilson. He was Secretary of Defense, and he actually had a pathological fear and hatred of the word research. [...] I decided therefore to use the word "programming". I wanted to get across the idea that this was dynamic, this was multistage, this was time-varying. I thought, let's [...] take a word that has an absolutely precise meaning, namely dynamic [...] it's impossible to use the word dynamic in a pejorative sense. Try thinking of some combination that will possibly give it a pejorative meaning. It's impossible. Thus, I thought dynamic programming was a good name. It was something not even a Congressman could object to. So I used it as an umbrella for my activities."

Richard Bellman, "Eye of the Hurricane: an autobiography" 1984.

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## Computing $D$ (1)

Obviously:

$$\begin{cases} D(i, 0) = i & \forall i \in \llbracket 1, n \rrbracket \\ D(0, j) = j & \forall j \in \llbracket 1, m \rrbracket \end{cases} \quad (1)$$

$\hookrightarrow D(i, 0) \rightarrow$  generate the empty string from the first  $i$  character of  $X$

$\hookrightarrow$  'best' way:  $i$  deletions

$\hookrightarrow$  recursive definition:  $D(i, 0) = D(i - 1, 0) + 1 \rightarrow$  considering a larger suffix, require one extra deletion

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## Computing $D$ (2)

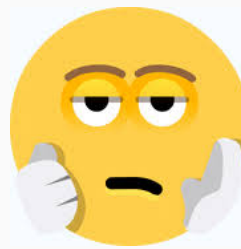
$$D(i, j) = \begin{cases} D(i - 1, j - 1) & \text{if } X[i] = Y[j] \\ \min \begin{cases} d[i - 1, j] + \omega_{\text{del}}(X[i]) \\ d[i, j - 1] + \omega_{\text{ins}}(Y[j]) \\ d[i - 1, j - 1] + \omega_{\text{sub}}(X[i], Y[j]) \end{cases} & \text{if } X[i] \neq Y[j] \end{cases} \quad (2)$$

$\hookrightarrow$  compute the edit distance **incrementally**

$\hookrightarrow$  at each step: best way to 'extend' the best solution achieved so far

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## For the moment...



$$\begin{cases} \omega_{\text{del}}(a) = 1 & \forall a \in \Sigma \\ \omega_{\text{sub}}(b) = 1 & \forall b \in \Sigma \\ \omega_{\text{ins}}(a, b) = 2 & \forall a, b \in \Sigma^2 \end{cases}$$

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## The edit distance table

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

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## The edit distance table

N										
O										
I										
T										
N										
E										
T										
N										
I										
#	0									
	#	E	X	E	C	U	T	I	O	N

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### The edit distance table

N										
O										
I										
T										
N										
E										
T										
N										
I	1									
#	0									
	#	E	X	E	C	U	T	I	O	N

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### The edit distance table

N										
O										
I										
T										
N										
E										
T										
N	2									
I	1									
#	0									
	#	E	X	E	C	U	T	I	O	N

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### The edit distance table

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

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### The edit distance table

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

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### The edit distance table

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

$X[i] \neq X[j] \Rightarrow \min \begin{cases} 1+1 & \text{insertion} \\ 1+1 & \text{deletion} \\ 0+2 & \text{substitution} \end{cases}$

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### 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

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## 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

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## Complexity



- time complexity:  $\mathcal{O}(n \times m)$
- space complexity:  $\mathcal{O}(n \times m)$

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## Are we done yet?



- efficient computation of the minimal edit distance **value**
- how can recover the corresponding sequence of editions?

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## Are we done yet?



- efficient computation of the minimal edit distance **value**
- how can recover the corresponding sequence of editions?  
⇒ **backtracking**

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## Backtracking

- for each entry in the edit distance table: keep track of the cell we came from  
i.e. the operation corresponding to the min

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

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## Weighted edit distance

Why would we add weights to the computation?

↔ e.g. different weights for different substitutions

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## Weighted edit distance

Spell checkers: some errors are more probable than other (even worst for T9)

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## Weighted edit distance

X	sub[X, Y] = Substitution of X (incorrect) for Y (correct)																													
	Y (correct)													Y (incorrect)																
	a	b	c	d	e	f	g	h	i	j	k	l	m	n	o	p	q	r	s	t	u	v	w	x	y	z				
a	0	0	7	1	3	4	0	0	2	1	18	0	1	0	0	3	7	6	0	0	1	3	5	9	9	0	1	0	5	0
b	0	0	9	9	2	2	3	1	0	0	0	5	11	5	0	10	0	0	2	1	0	0	8	0	0	0	0	0	0	0
c	6	5	0	16	0	9	5	0	0	0	1	0	7	9	1	10	2	5	39	40	1	3	7	1	1	0	0	0	0	0
d	1	10	13	0	12	0	5	5	0	0	2	3	7	3	0	1	0	43	30	22	0	0	4	0	2	0	0	0	0	0
e	388	0	3	11	0	2	2	0	89	0	0	3	0	5	93	0	0	14	12	6	15	0	1	0	18	0	0	0	0	0
f	0	15	0	3	1	0	5	2	0	0	0	3	4	1	0	0	0	6	4	12	0	0	2	0	0	0	0	0	0	0
g	4	1	11	9	2	0	0	0	0	0	1	3	0	0	2	1	3	5	13	21	0	0	1	0	3	0	0	0	0	0
h	1	8	0	3	0	0	0	0	0	0	2	0	12	14	2	3	0	3	1	11	0	0	2	0	0	0	0	0	0	0
i	103	0	0	0	146	0	1	0	0	0	0	6	0	0	49	0	0	0	2	1	47	0	2	1	15	0	0	0	0	0
j	0	1	1	9	0	0	1	0	0	0	0	2	1	0	0	0	0	5	0	0	0	0	0	0	0	0	0	0	0	0
k	1	2	8	4	1	1	2	5	0	0	0	0	5	0	2	0	0	6	0	0	0	0	4	0	0	0	0	0	0	0
l	2	10	1	4	0	4	5	6	13	0	1	0	0	14	2	5	0	11	10	2	0	0	0	0	0	0	0	0	0	0
m	1	3	7	8	0	2	0	6	0	0	4	4	0	180	0	6	0	9	15	13	3	2	2	3	0	0	0	0	0	0
n	2	7	6	5	3	0	1	19	1	0	4	35	78	0	0	7	0	28	5	7	0	0	1	2	0	0	0	0	0	0
o	91	1	1	3	116	0	0	0	25	0	2	0	0	0	0	14	0	2	4	14	39	0	0	0	18	0	0	0	0	0
p	0	11	2	0	6	5	0	2	9	0	2	7	6	15	0	0	1	3	6	0	4	1	0	0	0	0	0	0	0	0
q	0	0	1	0	0	0	27	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
r	0	14	0	30	12	2	2	8	2	0	5	8	4	20	1	14	0	0	12	22	4	0	0	1	0	0	0	0	0	0
s	11	8	27	33	35	4	0	1	0	1	0	27	0	6	1	7	0	14	0	15	0	0	5	3	20	1	0	0	0	0
t	3	4	9	42	7	5	19	5	0	1	0	14	9	5	5	6	0	11	37	0	0	2	19	0	7	6	0	0	0	0
u	20	0	0	0	44	0	0	0	64	0	0	0	0	2	43	0	0	4	0	0	0	0	2	0	8	0	0	0	0	0
v	0	0	7	0	0	3	0	0	0	0	0	1	0	0	1	0	0	8	3	0	0	0	0	0	0	0	0	0	0	0
w	2	2	1	0	1	0	0	2	0	0	1	0	0	0	0	7	0	6	3	3	1	0	0	0	0	0	0	0	0	0
x	0	0	0	2	0	0	0	0	0	0	0	0	0	0	0	0	0	9	0	0	0	0	0	0	0	0	0	0	0	0
y	0	0	2	0	15	0	1	7	15	0	0	2	0	6	1	0	7	36	8	5	0	0	1	0	0	0	0	0	0	0
z	0	0	0	7	0	0	0	0	0	0	0	7	5	0	0	0	0	2	21	3	0	0	0	0	0	3	0	0	0	0

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## Weighted edit distance



↔ use distance between letters on a keyboard

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