

## Why compute minimum edit distance?

Sometimes we want to know how “similar” two strings are.

- Could indicate morphological relationships:

walk - walks, sleep - slept

- Or possible spelling errors (and corrections):

definition - defintion, separate - seperate

- Also used in other fields, e.g., bioinformatics (gene sequences):

ACCGTA - ACCGATA

## Minimum edit distance: worked example

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15 September 2017



## MED is (one) way to measure similarity

- How many changes needed to go from string  $s_1 \rightarrow s_2$ ?

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

- To solve the problem, we need to find the best **alignment** between the words.

– Could be several equally good alignments.

## 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

## 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 T A L - L -
s d | i | i d      d d s s i | i
T - A B L E -      - - T A B L E
  
```

## How to find an optimal alignment

Brute force: Consider all possibilities, score each one, pick best.

How many possibilities must we consider?

- First character could align to any of:

```

- - - - - T A B L E -
  
```

- Next character can align anywhere to its right
- And so on... the number of alignments grows exponentially with the length of the sequences.

Maybe not such a good method...

## A better idea

Instead we will use a **dynamic programming** algorithm.

- Other DP (or **memoization**) algorithms we'll see later: Viterbi, CKY.
- Used to solve problems where brute force ends up **recomputing** the same information many times.
- Instead, we
  - Compute the solution to each subproblem **once**,
  - Store (memoize) the solution, and
  - Build up solutions to larger computations by combining the pre-computed parts.
- Strings of length  $n$  and  $m$  require  $O(mn)$  time and  $O(mn)$  space.

## Intuition

- Minimum distance  $D(\text{stall}, \text{table})$  must be the minimum of of:
  - $D(\text{stall}, \text{tabl}) + \text{cost}(\text{ins})$
  - $D(\text{stal}, \text{table}) + \text{cost}(\text{del})$
  - $D(\text{stal}, \text{tabl}) + \text{cost}(\text{sub})$
- Similarly for the smaller subproblems
- So proceed as follows:
  - solve smallest subproblems first
  - store solutions in a table (chart)
  - use these to solve and store larger subproblems until we get the full solution

## A note about costs

- Our first example had  $\text{cost}(\text{ins}) = \text{cost}(\text{del}) = \text{cost}(\text{sub}) = 1$ .
- But we can choose whatever costs we want. They can even depend on the particular characters involved.
  - Ex: choose  $\text{cost}(\text{sub}(c, c'))$  to be  $P(c'|c)$ , the probability of someone accidentally typing  $c'$  when they meant to type  $c$ .
  - Then we end up computing the most probable sequence of typos that would change one word to the other.
- In the following example, we'll assume  $\text{cost}(\text{ins}) = \text{cost}(\text{del}) = 1$  and  $\text{cost}(\text{sub}) = 2$ .

### Chart: starting point

		T	A	B	L	E
	0					
S						
T						
A						
L						
L						

- Chart $[i, j]$  stores two things:
  - $D(\text{stall}[0..i], \text{table}[0..j])$ : the MED of substrings of length  $i, j$
  - **Backpointer(s)** showing which sub-alignment(s) was/were extended to create this one.

### Filling first cell

		T	A	B	L	E
	0					
S	↑1					
T						
A						
L						
L						

- Moving down in chart: means we had a **deletion** (of S).
- That is, we've aligned (S) with (-).
- Add cost of deletion (1) and backpointer.

## Rest of first column

		T	A	B	L	E
	0					
S	↑1					
T	↑2					
A						
L						
L						

- Each move down first column means another deletion.
  - $D(ST, -) = D(S, -) + \text{cost}(\text{del})$

## Rest of first column

		T	A	B	L	E
	0					
S	↑1					
T	↑2					
A	↑3					
L	↑4					
L	↑5					

- Each move down first column means another deletion.
  - $D(ST, -) = D(S, -) + \text{cost}(\text{del})$
  - $D(STA, -) = D(ST, -) + \text{cost}(\text{del})$
  - etc

## Start of second column: insertion

		T	A	B	L	E
	0	←1				
S	↑1					
T	↑2					
A	↑3					
L	↑4					
L	↑5					

- Moving right in chart (from [0,0]): means we had an **insertion**.
- That is, we've aligned (-) with (T).
- Add cost of insertion (1) and backpointer.

## Substitution

		T	A	B	L	E
	0	←1				
S	↑1	↖2				
T	↑2					
A	↑3					
L	↑4					
L	↑5					

- Moving down and right: either a **substitution** or **identity**.
- Here, a substitution: we aligned (S) to (T), so cost is 2.
- For identity (align letter to itself), cost is 0.

## Multiple paths

		T	A	B	L	E
	0	←1				
S	↑1	↖↑2				
T	↑2					
A	↑3					
L	↑4					
L	↑5					

- However, we also need to consider other ways to get to this cell:
  - Move **down** from [0,1]: deletion of S, total cost is  $D(-, T) + \text{cost}(\text{del}) = 2$ .
  - Same cost, but add a new backpointer.

## Multiple paths

		T	A	B	L	E
	0	←1				
S	↑1	←↖↑2				
T	↑2					
A	↑3					
L	↑4					
L	↑5					

- However, we also need to consider other ways to get to this cell:
  - Move **right** from [1,0]: insertion of T, total cost is  $D(S, -) + \text{cost}(\text{ins}) = 2$ .
  - Same cost, but add a new backpointer.

## Single best path

		T	A	B	L	E
	0	←1				
S	↑1	←↖↑2				
T	↑2	↖1				
A	↑3					
L	↑4					
L	↑5					

- Now compute  $D(ST, T)$ . Take the min of three possibilities:
  - $D(ST, -) + \text{cost}(\text{ins}) = 2 + 1 = 3$ .
  - $D(S, T) + \text{cost}(\text{del}) = 2 + 1 = 3$ .
  - $D(S, -) + \text{cost}(\text{ident}) = 1 + 0 = 1$ .

## Final completed chart

		T	A	B	L	E
	0	←1	←2	←3	←4	←5
S	↑1	←↖↑2	←↖↑3	←↖↑4	←↖↑5	←↖↑6
T	↑2	↖1	←2	←3	←4	←5
A	↑3	↑2	↖1	←2	←3	←4
L	↑4	↑3	↑2	←↖↑3	↖2	←3
L	↑5	↑4	↑3	←↖↑4	↖↑3	←↖↑4

- Follow the backpointers to find the best alignment(s). This path, for example, corresponds to:
 

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

## Exercises

- Choose a different path through the backpointers and reconstruct its alignment.
- How many different optimal alignments are there?
- Redo the chart with all costs = 1 (Levenshtein distance), or some other set of costs, or using a different word pair.