# Lecture 19: Edit Distance 

Agenda:

- Scoring schemes in sequence comparison
- Edit distance
- Affine gap penalty scoring scheme

Reading:

- No textbook pages


## LCS problem review:

Definitions: - Sequence or String: dynamicprogramming is a sequence over the English alphabet

- Base/letter/character
- Subsequence:
the given sequence with zero or more bases left out e.g., dog is a subsequence of dynamicprogramming WARNing: bases appear in the same order, but not necessarily consecutive
- Common subsequence
- LCS problem: given two sequences $X=x_{1} x_{2} \ldots x_{n}$ and $Y=y_{1} y_{2} \ldots y_{m}$, find a maximum-length common subsequence of them.
- The LCS problem has the "optimal substructure" ...
- if $x_{n}$ is NOT in the LCS (to be computed), then we only need to compute an LCS of $x_{1} x_{2} \ldots x_{n-1}$ and $y_{1} y_{2} \ldots y_{m}$ ...
- similarly, if $y_{m}$ is NOT in the LCS (to be computed), then we only need to compute an LCS of $x_{1} x_{2} \ldots x_{n}$ and $y_{1} y_{2} \ldots y_{m-1} \ldots$
- if $x_{n}$ and $y_{m}$ are both in the LCS (to be computed), then $x_{n}=y_{m}$ and we need to compute an LCS of $x_{1} x_{2} \ldots x_{n-1}$ and $y_{1} y_{2} \ldots y_{m-1}$;
and then adding $x_{n}$ to the end to form an LCS for the original problem


## Sequence Alignment:

Definition: An alignment of two sequences $S_{1}$ and $S_{2}$ is obtained by first inserting spaces, either into or at the ends of $S_{1}$ and $S_{2}$, and then placing the two resultant sequences one above the other so that every character or space in either sequence is opposite a unique character or a unique space in the other sequence.

- An example, $S_{1}=$ rests, $S_{2}=$ stress


Note: space - is not allowed to be opposite to space -!!!

- Scoring scheme:

For every pair of characters in $\Sigma \cup\{-\}$, say $a$ and $b$, define a score $s(a, b)$ for them to be aligned in one column of the alignment.

- An example scoring scheme - LCS:
$s(a, a)=1$, for all $a \in \Sigma$; otherwise $s(a, \cdot)=0$
- Another notion: distance - how much it costs if $a$ is replaced by $b$ ?
- A distance measure (metric) must satisfy 3 conditions:

1. $d(a, a)=0$;
2. $d(a, b)=d(b, a)$;
3. $d(a, b) \leq d(a, c)+d(b, c)$.

## Edit Distance:

- A distance metric which specifies how much it costs to replace letter $a$ by letter $b-d(a, b)$.
- Goal: compute an edit transcript which minimizes the overall cost.
- Again, Edit Distance possesses the optimal substructure ... Explain how ???
Letting $\operatorname{Edit}[i, j]$ to denote the minimum cost of editing $S_{1}[1 . . i]$ into $S_{2}[1 . . j]$, then we have the following recurrence:

$$
\operatorname{Edit}[i, j]=\min \left\{\begin{array}{l}
\operatorname{Edit}[i-1, j]+d\left(S_{1}[i],-\right), \\
\operatorname{Edit}[i, j-1]+d\left(-, S_{2}[j]\right), \\
\operatorname{Edit}[i-1, j-1]+d\left(S_{1}[i], S_{2}[j]\right)
\end{array}\right.
$$

- Base cases ???


## Edit Distance:

- Pseudocode to implement the above recurrence
- Correctness
- Can return an associated Edit Transcript ... trace back
- Running time: $\Theta(n \times m)$

There are $n \times m$ entries each takes constant time to compute.

- Space requirement $\ldots \Theta(n \times m)$

Can be reduced to $\Theta(\min \{n, m\})$

## Scoring Schemes:

- Edit distance:

1. letter dependent scoring scheme;
2. letter independent scoring scheme: match, mismatch, insertion/deletion (indel)

- An edit transcript $\Longleftrightarrow$ an alignment

Score/Cost of the alignment is the sum of scores/costs of columns ...

- Now ask: from rests to stress,

Are r and e deleted separately, or they are deleted at the same time?

If deleted at the same time, how do we assign a cost for it?

- Or, consecutive spaces should be counted as a gap ...
- Affine gap penalty scoring schemes:
penalties for a gap: gap opening $d_{o}$ and gap extension $d_{e}$
- Now how do we compute an optimal edit transcript? Consider three cases ...


## Edit Distance with Affine Gap Penalty Scoring Scheme:

- It still possesses the optimal substructure ...

Letting $\operatorname{Edit}_{M}[i, j]$ to denote the minimum cost of editing $S_{1}[1 . . i]$ into $S_{2}[1 . . j]$ where the last operation is either a match or a mismatch;
Letting $E \operatorname{dit}_{I}[i, j]$ to denote the minimum cost of editing $S_{1}[1 . . i]$ into $S_{2}[1 . . j]$ where the last operation is an insertion;
Letting $E_{i t_{D}}[i, j]$ to denote the minimum cost of editing $S_{1}[1 . . i]$ into $S_{2}[1 . . j]$ where the last operation is a deletion.

- Recurrence:

$$
\widetilde{\operatorname{Edit}}[i, j]=\min \left\{E \operatorname{dit}_{M}[i, j], E \operatorname{dit}_{I}[i, j], E d i t_{D}[i, j]\right\}
$$

$$
\begin{aligned}
& \operatorname{Edit}_{M}[i, j]=\widetilde{\operatorname{Edit}}[i-1, j-1]+d\left(S_{1}[i], S_{2}[j]\right) ; \\
& \operatorname{Edit}_{I}[i, j]=\min \left\{\begin{array}{l}
\operatorname{Edit}_{M}[i, j-1]+d_{o}+d_{e}, \\
\operatorname{Edit}_{I}[i, j-1]+d_{e} \\
\operatorname{Edit}_{D}[i, j-1]+d_{o}+d_{e}
\end{array}\right. \\
& \operatorname{Edit}_{D}[i, j]=\min \left\{\begin{array}{l}
\operatorname{Edit}_{M}[i-1, j]+d_{o}+d_{e} \\
\operatorname{Edit}_{I}[i-1, j]+d_{o}+d_{e} \\
\operatorname{Edit}_{D}[i-1, j]+d_{e}
\end{array}\right.
\end{aligned}
$$

Output $\widetilde{\operatorname{Edit}}[n, m]$ !

- Base cases ???
- Running time? Space complexity?

Have you understood the lecture contents?

| well | ok | not-at-all | topic |
| :--- | :--- | :--- | :--- |
| $\square$ | $\square$ | $\square$ | sequence alignment |
| $\square$ | $\square$ | $\square$ | edit distance |
| $\square$ | $\square$ | $\square$ | DP for edit distance |
| $\square$ | $\square$ | $\square$ | affine gap penalty scoring scheme |
| $\square$ | $\square$ | $\square$ | edit distance with AGPSS |
| $\square$ | $\square$ | $\square$ | DP for edit distance with AGPSS |

