Levenshtein Distance Nick Sebasco

Definition

The Levenshtein distance, also known as the minimum edit distance is a way of assessing how different two strings are.

Strings are a data structure (โครงสร้าง ข้อมูล) defined as an order collection of characters. In many programming languages the sequence of characters is enclosed with quotes.



Actual	OCR
1AY4DF	LAYADF



Motivation

In euclidean space there are straight forward distance metrics that look at how similar the components making up two vectors are. For instance, the **Manhattan** distance:

$$D_{manhattan} = \sum |x_i - y_i|$$

$$D_{manhattan} = |1 - 0| + |2 - 0| + |0 + 1| = 4$$

It is tempting to try a similar strategy for strings. Let's compute a string distance which adds 1 to the distance every time two characters at the same position are different and a 0 each time they are the same. Clearly more similar strings will have smaller distances.

$$D_{\text{string}} = \Sigma 1 \text{ if } x_i = y_i \text{ else } 0$$

Levenshtein Distance

```
ีน = "กาม"
x = "กอออ"
y = "ามา"
D_{ux} = 1) Edit \mathbf{a} \rightarrow 1 2) Edit \mathbf{a} \rightarrow \mathbf{a} 3) Delete \mathbf{a} = 3
D<sub>uv</sub> = 1) Insert ก = กามา 2) Delete า = กาม = 2
```

 $D_{uy} < D_{ux}$ **u** more similar to **y**.

Is that the best we can do?

The string distance metric in the last example seems to miss the fact that **y** and **u** have more similar characters that **x** and **u**. Also **y** is the same length as **u**. Thus **y** and **u** appear to be more similar. If we consider the more general question of transforming one string into another and restrict ourselves to the operations **add**, **delete**, and **insert**, then we can create a more robust string similarity metric.

String	Operation	Position	Result
ถนน	delete	1	นน
นก	Insert,insert,insert	1,2,3	รังนก
มะนาว	มะนาว edit		มะตาว

Algorithm

- Given two strings u and v of length m and n respectively. Initialize an m+1xn+1 matrix M.
- 2. Let \mathbf{r}_i be the \mathbf{i}^{th} row of \mathbf{M} and \mathbf{c}_i be the \mathbf{i}^{th} column of \mathbf{M} . Then $\mathbf{r}_1 = [0, ..., \mathbf{m}]$ and

$$c_1^T = [0, ..., n]$$

- 3. Iterate over **u** from $i = 1 \dots m$ and iterate over **v** from $j = 1 \dots n$.
- 4. if $\mathbf{u}[\mathbf{i}] = \mathbf{v}[\mathbf{j}]$ then cost = 0 else $\mathbf{u}[\mathbf{i}] = -\mathbf{v}[\mathbf{j}]$ then cost = 1
- 5. Set cell **M[i, j]** of the matrix equal to the minimum of:
 - I. The cell (one row above): M[i-1,j] + 1.
 - II. The cell (one column to the left) 1: **M[i,j-1]** + 1.
 - III. The cell (one row above and one column to the left): **M[i-1,j-1]** + cost.
- 6. After the matrix has been filled, the final value in the matrix will be the distance!

bed

lead

		1	е	а	d
	0	1	2	3	4
b	1	1	2	3	4
е	2	2	1	2	3
d	3	3	2	2	2

Figure: Computing the Levenshtein distance

Python Implementation

```
import <u>numpy</u> as <u>np</u>
def lev(s1: str, s2: str) -> np.matrix:
    '''AKA: Edit distance
    levenshtein distance between two sequences s1, s2 is the minimum number of single
    element (insert/ delete/ substitution) required to transform s1 -> s2.
    # 1. initialize matrix
    T = \underline{np}.zeros((len(s1) + 1, len(s2) + 1))
    for i in range(max(len(s1)+1, len(s2)+1)):
        if i < len(s1)+1:
            T[i, 0] = i
        if i < len(s2)+1:
            T[0, i] = i
    # 2.
    for i in range(len(s1)):
        for j in range(len(s2)):
            T[i + 1, j + 1] = min(T[i, j + 1], T[i + 1, j], T[i, j]) + 1 \text{ if } s1[i] != s2[j] \text{ else } T[i, j]
    return T
print(lev("bed","lead"))
```

Application: Covid-19 Variants

The string alpha represents a random snippet of RNA base pairs from the Covid-19 alpha variant, see the reference below for a link to the full genome [1]. Let mutant₁ and mutant₂ represent disparate mutations of the alpha variant. Use the levenshtein distance to calculate which mutation is most similar.

Variant	Base Pairs
alpha	attaaaggtt tataccttcc
mutant ₁	caggatagctt atacctaacc
mutant ₂	ctaaaagtt tatagttcc

[1] Covid-19 Alpha variant genome: https://www.ncbi.nlm.nih.gov/nuccore/MN908947.3