

ADVANCED DATA MANAGEMENT

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What will be discussed today

- Distributed Job Scheduling
(actual implementation)
- General introduction to NLP

What is NLP

NLP is an interdisciplinary field concerned with the interactions between computers and natural human languages (e.g. English) — speech or text

NLP – The classics

The first way to approach the problem, is to have a system to evaluate how much similar two strings are. It can be done with two methods:

- Syntactic
- Semantic

NLP – The classics → Syntactic

- Hamming distance: it is evaluated on two strings of equal length, it is the number of characters with the same index which differ
- Levenshtein distance: it is evaluated between two strings, it is the number of edits required to change one sequence to another. Operations are: insertion, deletion and substitution
- Problem: position is significant

NLP – The classics → Syntactic

- Hamming distance: very simple algorithm, efficient ($O(N)$), but can be employed in a limited number of cases
- Levenshtein distance: more complex than hamming, can be used on every string, complexity is ($O(N^2)$) with dynamic programming

NLP – The classics → Syntactic

Algorithm Edit distance

Input: $\alpha = \alpha_1 \dots \alpha_n$ and $\beta = \beta_1 \dots \beta_m$

```
1: for  $i \leftarrow 0$  to  $n$  do
2:    $D_{i,0} \leftarrow i$ ;
3: end for
4: for  $j \leftarrow 0$  to  $m$  do
5:    $D_{0,j} \leftarrow j$ ;
6: end for
7: for  $i \leftarrow 1$  to  $n$  do
8:   for  $j \leftarrow 1$  to  $m$  do
9:      $t \leftarrow (\alpha_i = \beta_j)? 0 : 1$ ;
10:     $D_{i,j} \leftarrow \min\{D_{i-1,j-1} + t, D_{i,j-1} + 1, D_{i-1,j} + 1\}$ ;
11:   end for
12: end for
13. return  $D_{n,m}$ 
```

NLP – Phonetic → Syntactic

- This class of algorithms tries to capture the pronunciation similarities of words
- Soundex: very old algorithm, used in some system for indexing strings
- Metaphone: improvement on Soundex and less old (1990 → 2009 last version). It has several variants. Shows better results than its predecessor
- Main limitation: each algorithm is designed to work well for a single language

NLP – Phonetic → Syntactic

```
1  function SOUNDEX(word)
2    result := upperCase(word1);
3    for i ∈ {2, ... , length(word)} do
4      code := {
5        1 if wordi ∈ {b, f, p, v},
6        2 if wordi ∈ {c, g, j, k, q, s, x, z},
7        3 if wordi ∈ {d, t},
8        4 if wordi ∈ {l},
9        5 if wordi ∈ {m, n},
10       6 if wordi ∈ {r},
11       ε otherwise;
12
13     if resultlength(result) ≠ code then
14       result := result ◦ code;
15
16     while length(result) < 3 do
17       result := result ◦ 0;
18
19     return result;
20  end.
```

NLP – Token based → Syntactic

- This class of algorithms tries is an adaptation of set similarity algorithms, thus they are express in mathematical terms

- $$J(A, B) = \frac{|A \cap B|}{|A \cup B|} = \frac{|A \cap B|}{|A| + |B| - |A \cap B|}$$

Jaccard index

- $$DSC = \frac{2|X \cap Y|}{|X| + |Y|}$$

Sorensen-Dice formulae

- Problem 1: what is an element of a set?
- Problem 2: repeated tokens do not matter

NLP – Syntactic techniques recap

- Classic techniques: good for matching a small string against a large corpus of text (e.g., spell checker). Very bad when the order of words is not meaningful
- Phonetic: good for indexing single terms (compact representation). They need to be developed for each language to function
- Token based: good for comparing sentences, specifically when ordering is not meaningful. They let the programmer identify what is a token(letter, triplet, n-gram, word). Do not take into account token frequencies

NLP – Information Retrieval (IR)

- Extensions of the token based methods in order to be able to reason about what is contained in a text
- Bag of Words (BoW): the most simple application of IR. It transforms text in a frequency set of the words found in it.
- Term Frequency – Inverse Document Frequency (TF-IDF): technique used to measure the importance of a term in a specific corpus of text which is part of a set of multiple documents

NLP – BoW

	about	bird	heard	is	the	word	you
About the bird, the bird, bird bird bird	1	5	0	0	2	0	0
You heard about the bird	1	1	1	0	1	0	1
The bird is the word	0	1	0	1	2	1	0

NLP – BoW (useful preprocessing)

- Remove punctuation → while useful to the semantic of a sentence, it serves no purpose in the syntax of it
- Remove stop words → very frequent and not meaningful words (e.g., “and” “or” “of”). They depend on the analyzed language
- Stemming → same word but with some derivation returned to its original form (e.g., playing → play)

NLP – TF-IDF

- Two element of the formula:
 - $TF(i,j) = n(i,j) / |d_j|$ → number of occurrences of term i in document j over the overall number of words in j
 - $IDF(i) = \log_{10}(|D| / |\{d: i \text{ in } d\}|)$ → logarithm with base 10 of the total number of documents over the number of documents that contains the term i
 - $TF-IDF(i,j) = TF(i,j) * IDF(i)$

NLP – TF-IDF - intuition

- TF: the more frequent a term is in a document, the more relevant it is. The divisor is used to avoid to favor longer documents over smaller ones
- IDF: it was proposed as first as an heuristic to measure the specificity of a term. If a lot of documents contains it, it is an indication of how much a term could be similar to a “stop-word” and thus not being relevant

NLP – BoW and TF-IDF problems

- While in some cases the order of the words is not meaningful, when trying understand “what” a document is about, this information is crucial
- All the techniques explained until now share one common problem: they have no way to actual understand the semantics of a text, since it has no information about the context of the retrieved terms