SPELL-CHECKING QUERIES BY COMBINING LEVENSHTEIN AND STOILOS DISTANCES

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Context

oi de neuf? Iropos de	Doc'CISMeF
	Outil de recherche en médecine
ipe	Rechercher Avancée
lications	☞ tous types et documents
rtenaires	C uniquement les recommandations professionnelles
ertissement	C uniquement les documents concernant l'enseignement C uniquement les documents et associations concernant les
de	patients
	Listes et index : alphabétique, thématique, types de ressources
(=79374) a 3 mai 2011	
Copyright 1995-2011	
Dublin Core	
Nous adhérons aux	Portail Terminologique de Santé
principes de la	Consulter le MeSH et les autres terminologies de santé
charte HUNcode.	Bechercher
	Mechercher
	Connexion S'inscrire

Catalog & Index of Health Resources in French on the Internet

CISMeF = quality controlled health gateway for French institutional health resources

Doc'CISMeF: a search tool

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to search within the catalog CISMeF more than 82,000 documents
specific of the health resources available on the Internet, such as association, patient information, community networks

3 types of users:

- Patients
- Students
- Clinicians

Introduction

- Increase in the number of users querying different search engines
- Internet became a major source of health information
- Medical vocabularies are difficult to handle by non-professionals
- Did you mean:" of Google or "Also try:" of Yahoo

Introduction

Purpose: Spelling correction for medical queries in French.

Method: Spelling correction based on comparing the query with a dictionary.

Tools: The string distance of Stoilos and the Levenshtein edit distance to correct spelling errors. We propose here to combine them.

String distance: Levenshtein

 Minimum number of edit operations (insertion, deletion, substitution) to transform one string into the other

String distance: Levenshtein

The Normalized Levenshtein (LevNorm) in the range [0, 1]

LevNorm
$$(c_1, c_2) = \frac{Lev(c_1, c_2)}{Max(length(c_1), length(c_2))}$$

□ <u>Example</u> :

LevNorm (Trigonocepahlie, Trigonocephalie) = 2/15 = 0.133 Lev(Trigonocepahlie, Trigonocephalie) = 2 max(length(Trigonocepahlie), length(Trigonocephalie)) = max(15,15) = 15

The similarity among two entities is related to their commonalities as well as to their differences. Thus, the similarity should be a function of both these features.

$$Sto(s_1, s_2) = Comm(s_1, s_2) - Diff(s_1, s_2) + Winkler(s_1, s_2)$$

The function of commonality computes the longest common substrings between 2 strings

$$Comm (s_1, s_2) = \frac{2 \times \sum_{i} length (MaxComSubString_i)}{length(s_1) + length(s_2)}$$

• Example: s_1 = 'Trigonocepahlie' et s_2 = 'Trigonocephalie' $length(MaxComSubString_1) = length(Trigonocep) = 10$ $length(MaxComSubString_2) = length(lie) = 3$

Comm(Trigonocepahlie, Trigonocephalie) = 13/15 = 0.866

Based on the length of the unmatched strings that have resulted from the initial matching step

$$Diff(s_1, s_2) = \frac{uLen_{s_1} \times uLen_{s_2}}{p + (1-p) \times (uLen_{s_1} + uLen_{s_2} - uLen_{s_1} \times uLen_{s_2})}$$

 s_1 = 'Trigonocepahlie' and s_2 = 'Trigonocephalie' and p = 0.6

 $uLen_{S1} = 2/15$ and $uLen_{S2} = 2/15$

So $Diff(s_1, s_2) = 10/787 = 0.0254$

The Winkler correction:

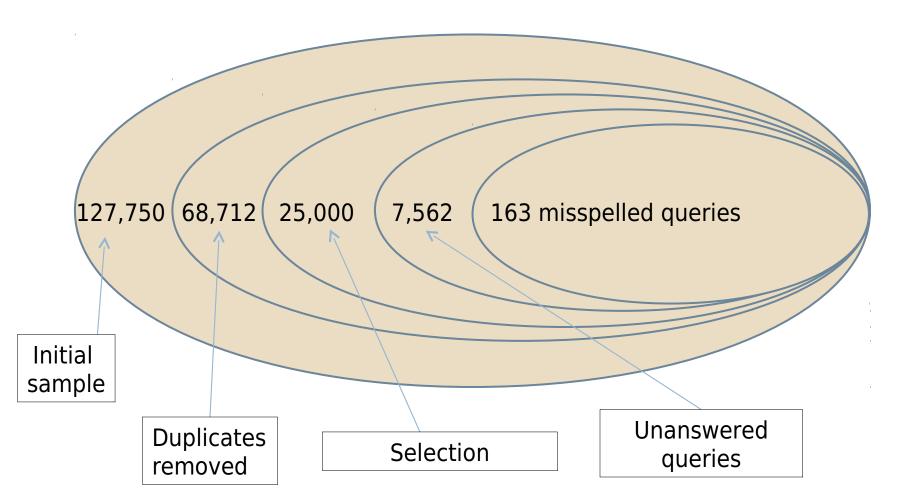
Winkler
$$(s_1, s_2) = L \times p' \times (1 - Comm(s_1, s_2))$$

 s_1 = 'Trigonocepahlie' and s_2 = 'Trigonocephalie' L = 4 and p' = 0.1So $Winkler(s_1, s_2) = 4/75 = 0.053$

Altogether

Sto(Trigonocepahlie, Trigonocephalie) = 13/15 - 10/787 + 4/75= 0.894

Materials: Queries



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Choice of thresholds

Levenshtein and Stoilos string distances require a choice of thresholds to obtain a manageable number of propositions of correction to the user. So we have tested this number for 163 misspelled queries.

	Method								
	Levenshtein			Stoilos			Levenshtein & Stoilos		
Thresholds	<0.2	<0.1	<0.05	>0.7	>0.8	>0.9	Lev < 0.2 Stoilos > 0.8	Lev < 0.2 Stoilos > 0.7	
Nb of answers	224 1.37	76 0.46	8 0.04	1454 8.92	489 3	140 0.85	179 1.09	213 1.30	

Evaluation

$$\frac{\text{Precision}}{\text{Queries correctly corrected}}$$

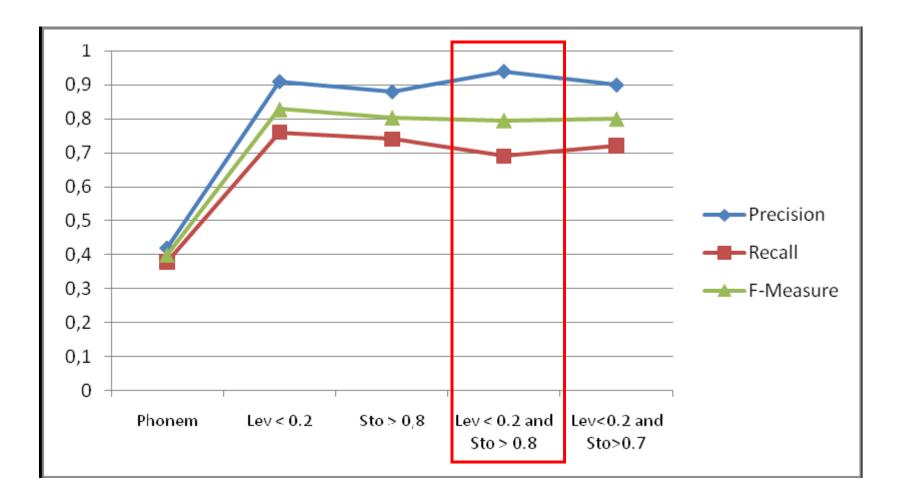
$$F-Measure = \frac{2 \times Precision \times Recall}{Precision + Recall}$$

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Results

Method	Recall	Precision	F-Measure
Phonetic transcription	0.38	0.42	0.399
Levenshtein < 0.2	0.76	0.91	0.8283
Stoilos > 0.8	0.74	0.88	0.8039
Levenshtein < 0.2 & Stoilos > 0.8	0.69	0.94	0.7958

Evaluation



Conclusion

A method to automatically correct misspelled queries submitted to health search tool

 The combination of the 2 distances gives a recall of 69% and a precision of 94%

This combination has increased the precision, but decreased the recall

The functionality is implemented in CISMeF

Perspectives

- Misspelled queries categorized according to their number of words
- The configuration of a keyboard, by studying the distances between keys