## 3 Sterrming and truncation

### 3.1 Introduction

One way of broadening searches in information retrieval is to use systematic abbreviation of words so as to bring together words which are morphologically related, in the hope that they will also be semantically related. This can be done manually or automatically.

Information retrieval intermediaries use manual truncation to conflate words which are both morphologically and semantically related. Intermediaries use their linguistic krowledge to avoid drawing in words which might seriously decrease the precision of the search. Truncation is often combined with boolean OR to bring in other symonyms. For example, the concept "communism" might be submitted to an IR system as "communis* OR marxis*" rather than as "commun* QR marx*" which would lead to the retrieval of records indexed under "communication" etc.

Manual truncation is not a particularly easy or natural skill to acquire, and cannot be considered for casual catalogue users. It is a facility which can be provided for those, such as library staff, who wish to Learn how to use it. Many of the "keyword" type online catalogue systems do allow for manual truncation [1].

The discussion here is Limited to processes of automatic abbreviation or truncation which aim to conflate related words by reducing them to their stems. The word-segments to be removed are referred to as affixes. An affix can be a suffix (Like "ation") or a prefix (Like "pre"). Many prefixes camnot safely be removed except in narrowly defined subject areas (such as chemical terminology), because they tend to have a more drastic effect on the meaning of a word than do suffixes, which are often inflexional.

The next three sections form a fairly condensed survey of some of the stemming techniques which have been published. They are of a rather technical nature. Readers who are primarily interested in online catalogues might prefer to skip to Section 3.5.

### 3.2 Methods and techniques in algorithm construction

Many algorithms have been reported. Some of the ways in which they differ are outlined in the following sections.

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a very high quality, the length of time taken often makes this method impractical. In their evaluation, they test a method (using the frequency of word endings) for the automatic generation of possible suffixes. They concluded that "fully automated methods perform as well as procedures which involve a Large degree of manual involvement in their development" (183).

### 3.2.5 Users' needs

The main function of conflation algorithms must be to improve recall; there will always be some searches where there is a loss of precision. The balance between recall and precision must be chosen to suit different classes of users. An industrial user of a retrieval system who needs a comprehensive search might be prepared to examine a substantial proportion of irrelevant material Ccaused by overstemming). For general library catalogue use, on the other hand, under-stemming is to be preferred to overstemming.

### 3.3 Conflation algorithms: a review

### 3.3.1 INTREX

One of the first conflation algorithms to be developed and tested was part of the Project INTREX Csee Overhage and Reintjes [4] for a general review). Lovins [5], who participated in this project, produced a list of suffixes by first examining a preliminary list generated from the endings of words from the Project INTREX catalogue. The list was used to see when the use of a given ending from the word in the dictionary would result in a mismatch, or in the omission of a stem which ought to match. This manual assessment allowed the author to refine the list of endings and to compile word specification and recoding rules. The final list contained around 260 suffixes. It was used in conjunction with both context rules and recoding rules.

### 3.3.2 RADCOL

Lowe and others [6] tested two algorithms as part of the RADCOL project. The first used two passes through a list of 95 suffixes; the second used a single pass longest match algorithm with a Longer List of 570 suffixes. After tests, the second algorithm was adopted. Lowe and his colleagues obtained this list by a multi-stage process. First the characters of the most frequent words in the index were reversed, and the reversed words were sorted in alphabetical order. Then a minimum string length was estabLished. The list was scanned for repeated character strings, on the assumption that strings which occurred with more than a certain frequency might be possible suffixes. Finally, these character strings were examined manually and
suffixes were selected from them. The comprehensiveness of this suffix list meant that the number of context and recoding rules could be reduced, increasing the simplicity of the algorittm.

### 3.3.3 Generation of suffix Lists

Lennon and his colleagues, in the course of their evaluation of conflation algorithms [3], extended the method used in the RADCOL Project and in the INSPEC project Cdiscussed below). A list of reversed words was used to produce a list of word endings which occurred with more than a certain frequency. These can be assumed to be suffixes al though if this algorithm is to be used automatically, no recoding is possible. This is a simple but unwieldy approach. Since it operates as a Longest match algorithm, the inclusion of a string "iveness" in a suffix dictionary also necessitates the inclusion of the substrings "veness", "eness", "ness" and so on. The proportion of strings which have a real utility is therefore reduced.

This method was also used by Tarry [7] to generate several sets of equifrequent character strings from the ends of words. The method involves selecting from a body of text character strings of variable length occurring with approximately equal frequencies and with low sequential dependence. This suffix generation procedure can be used for automatically determining subject-specific or languagespecific lists of suffixes. The incentive for using this technique for suffix generation was the supposition that character strings representing suffixes would occur more frequently than other terminal character strings. As well as this, it has been observed that letter dependency within words decreases at the boundaries of word units such as affixes. Tarry's algorithm works on the Longest match principle and has no restriction on suffix removal other than that the remaining stem should be of a minimum length. Since there is no restriction on removal the algorithm is context-free and uses neither recoding nor partial matching. Tarry justifies this approach by the desirability of eliminating "the Large amount of manual preprocessing required, both in the construction of the suffix Lists, and in the formulation of the suffix removal rules" [7, p21]. This algorithm was compared with the INSPEC algorithm; retrieval tests with the Cranfield 1400 test collection were made and it was found that the algorithm performed at least as well as the traditional algorithm [7, p78].

### 3.3.4 INSPEC

A conflation algorithm was designed by Field [8] at INSPE[ with British Library funding. The List of suffixes was compiled manually after consulting a Key Letter In Context (KLIC) index. This algorithm uses a mixture of longest
match and iterative suffix removal and incorporates several features which were designed to improve its effectiveness: minimum stem length, recoding rules and three stage conflation. This last application is particularly interesting. The word to be conflated is first dealt with by Algorithm 0 which removes stop words and common endings such as plural forms (this stage is partially iterative). Words which are not stopped are then treated by Algorithm 1 which removes all other suffixes which are present in a Longest match routine. In a final stage, Algorithm 2 makes adjustments to the stem, usually on the basis of stem length. Field claims that this use of a three stage process increases the overall efficiency.

### 3.3.5 Stemming in SMART and FIRST

The IR systems used in the SMRRT projects incorporated stemming. The SMART system bases all dictionaries on word stems rather than original words. The suffixes which generate the word stems are Listed in a suffix dictionary, and each one carries one or more syntactic codes. These must be matched with complementing codes attached to the word stems in order to determine which suffixes match which stems [9, p32].

Dattola [10] has described FIRST - the Flexible Information Retrieval System for Text - which is based on the methods developed during the SMART project. The most important part of this procedure is a stem dictionary; this is the basis of the conflation procedure. Words are added to the stem dictionary if they fail to match an existing stem and are more than three characters long. This method uses a stem dictionary of whole words rather than actual stems; new words are not added to the stem dictionary if they are suffix variations of existing stem entries.

### 3.3.6 MORPHS

Bell and Jones have described the retrieval system MORPHS in a number of articles including [11]. This system (the name means "Minicomputer Dperated Retrieval (Partially Heuristic) System") is used at the Malaysian Rubber Producers' Research Association. It incorporates automatic stemming. Bell and Jones [12] discussed the use of roles and stemming in an earlier version of the system as a means of improving recall and also incorporating some syntactic knowledge. They believed that the two techniques could be combined by replacing the suffixes by a limited number of role indicators. In this system stemming was performed manually by the searcher who could either, as in their example, search for MIX; or MIX (role A) - to include MIXING; or MIX (role D) - to include MIXED. An extensive suffix list is used; its size is increased by its treatment of exceptions (the stems "cation" and "station" are included and are used in preference to the stem "ion") and by
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MARS has morpheme dictionaries and grammar rules for each language. They are used to split words into prefix, stem, derivational and inflectional elements. The extracted word stems are collected in a stem-file in which pointers back to the textwords containing the particular stem can be followed, enabling retrieval of these words. The morpheme dictionary contains affixes, inflectional endings and filLers. Each entry is stored with a 32-bit string indicating special morpheme characteristics and certain compositional properties. The morphemes in the dictionary are the Longest possible strings obtainable from all of the possible derivations ("traditionality" for example would be viewed as a derivation of "tradition" and not "trad(e)"). This morpheme dictionary is supplemented by two smaller Lists. One includes "irregular" stems Like Latin and Greek plurals and irregular verb forms. The other list contains strings which regularly undergo graphemic change Clike "y" to "ie"); these transformations are processed automatically. A pre-processor checks to see if string tramsformations are necessary. After this, the three lists are used by a decomposition grammar which deals with each word. After having reached a certain stage in a word Ca prefix for example) certain conditions have to be fulfilled if the word is to be passed to the next stage. These conditions are listed in the morpheme grammar for the Language.

MARS was tested by a retrieval expert who carried out twelve real searches, once with and once without MARS. Recall was increased by 68\% when MARS was used. Moreover, this was achieved without a significant decrease in precision (this did decrease but only by $7 \%$ from 68\% to 61\%). There were difficulties with compound words and phrases and with verbs; these were caused by Limitations within the structure of MARS and can be offset by modifications to the search strategy used.

### 3.3.9 Porter

An iterative algorithm was developed by Martin Porter [2] at the University of Cambridge Computer Laboratory. He uses a concept which he calls the "measure" of a word. This is the number of vowel-consonant transitions in the word. It is used in some of the conditional rules: for example "remove terminal 'ance' if the measure is greater than one". The algorithm is a five-step, partially iterative procedure using a dictionary of around 60 suffixes. Porter notes that a point is reached in the development of a conflation algorithm when the inclusion of additional rules to improve performance in one area leads to a corresponding decrease in performance in another area. He warns that unless this tendency is guarded against it is very easy for the algorithm to become more complicated than
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### 3.4 Evaluating conflation algorithms

The effectiveness of stemming algorithms can be evaluated by assessing the degree to which 'erms are overstemmed and understemmed. One measure is the proportionate decrease in the number of distinct terms after stemming.

Lennon and others tested five conflation algorithms [3] as part of an evaluative study. They confirmed a previous suggestion by Landauer and Mah 1980 [18] that the the RADCOL algorithm tended to overstem (reducing "posed", "positively" and "positioning" to "pos"). The Porter algorithm tended to understem (reducing "accuracy" to "accurac" but "accurate" and "accurately" to "accur").

These conclusions are supported by the compression results which were achieved by Lennon and his colleagues. With the Brown Corpus, Porter achieved the Least compression (38.8\%) and RADCOL achieved the greatest (49.1\%). The other algorithms tested achieved $45.5 \%$ (Lovins) and $47.5 \%$ (INSPEC). Several test databases were used and while the percentage compression achieved did vary significantly according to database, the relative compression achieved by different algorithms was similar. Retrieval tests demonstrated that algorithms which tended to stem generously did not necessarily increase retrieval effectiveness; the Porter algorithm tended to understem, but it performed better in the test than the RADCOL algorithm which tended to overstem. The INSPEC algorithm, on the other hand, is also a strong algorithm, but this gave the best precision orientated search. Lennon and his colleagues also performed a test for recall effectiveness. In this test, a similarity measure using trigrams performed well; but the Porter algorithm performed as effectively. They conclude that "... there is no relationship between the strength of an algorithm and the consequent retrieval effectiveness arising from its use".

Altering the emphasis slightly, significance tests showed that mone of the conflation algorithms tested was significantly worse, and several were significantly better, than use of unstemmed words.

### 3.5 Stemming in online catalogues

As mentioned in 2.4.1, we do not know of any catalogue accessing a general collection which uses automatic stemming.

Among specialised or experimental catalogues, there is CITE, which uses a stemming procedure designed for medical terminology [19]. For the intermediate version of Okapi we used a slightly modified version of Porter's algorithm [2]. This system was not put out for live use, but experiments involving the repetition of real searches from transaction
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