A New Multi-Phase Algorithm for Stemming in Farsi Language Based on Morphology

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Abstract—The main goal of stemming is to standardize words by reducing a word to its origin. In this paper a new algorithm for stemming in Farsi (Persian) language is presented. This stemmer is based on removing the suffixes and prefixes, and a database is used for saving the exceptions to decrease error rate. In the proposed method the speed of stemmer and also the percentage of errors are improved. The evaluation results on the prototype document collections show significant improvement in precision and recall in comparison with other well-known methods.

Index Terms—Farsi, persian, language, stemming.

I. INTRODUCTION

Stemming is a fundamental step in processing textual data preceding the tasks of information retrieval, text mining, and natural language processing. The common goal of stemming is to standardize words by reducing a word to its base. In languages with very little inflection such as English and Mandarin Chinese, the stem is usually not distinct from the “normal” form of the word. However, in other languages, stems are more noticeable [1]. For example, the English verb stem end is indistinguishable from its present tense (except in the third person singular) [2]. There is much research of the effects of stemming on searches of English document collections [3]. Stemmers such as the Lovins and Porter stemmers sometimes improve precision/recall scores [4]. However, they only stem English terms.

Farsi or Persian is an Indo-European language, spoken and written primarily in Iran, Afghanistan, and a part of Tajikistan. Like English, Farsi has affinitive morphology. In other words, suffixes and prefixes are concatenated to words to modify meaning. Farsi is read from right to left, so that prefixes are attached to the root, and suffixes are attached to the left. Like English nouns, Farsi nouns are modified to signify possession, agency and plurality. However, Farsi verbs are modified more extensively than English verbs. Farsi verb forms vary according to tense, person, negation, and mood. To facilitate the information retrieval in Farsi search and display technology project [5], Kazem Taghva, Russell Beckley, and Mohammad Sadeh designed and implemented a Farsi language stemmer [3]. Its aim was to stem a word to find a more general form of it, possibly its root. For example, stemming the term interesting may produce the term interest or “interes”. Though a stemmer might not always give the root, that algorithm want all words that have the same stem to have the same root. On the other hand, for information retrieval, that stemmer do not always want all words with a given root to have the same stem because some words with the same root may be topically uncorrelated e.g. preside and president.

In this paper a Farsi algorithm which is based on morphology is described (like porter algorithm in English). The algorithm is implemented and its problems were found. So these problems were solved by presenting an improved algorithm. Finally the results of first algorithm and improved algorithm were compared. The results of improved algorithm were better.

The paper is organized as follows: Section two describes related works have been done in this field. In Section three Persian morphology is described and in Section four our new Farsi stemming algorithm is proposed. Section five describes Experimental results and discussion. Finally in Section six conclusion and future works are outlined.

II. RELATED WORKS

Most stemming approaches are based on the target languages morphological rules (e.g., the Porter stemmer for the English language [6]) where suffix removal is also controlled by quantitative restrictions (e.g., ‘ing’ is removed when the resulting stem has more than three letters as in “jumping,” but not in “king”) or qualitative restrictions (e.g., ‘-ize’ is removed if the resulting stem does not end with ‘e’ as in “seize”). Certain ad hoc spelling correction rules can also be applied to improve conflation accuracy (e.g., “running” gives “run” and not “runn”), particularly when phonetic rules are applied to facilitate easier pronunciation. Another approach consults an online dictionary to obtain better conflation results [7], while Xu & Croft suggest a corpus-based approach that more closely reflects the language use rather than all its grammatical rules [8]. Few stemming procedures have been suggested for languages other than English. The proposed stemmers usually pertain to the most popular languages and some of them, like the Finnish language [9], seem to require a deeper morphological analysis to achieve good retrieval performance [10]. Algorithmic stemmer ignores word meanings and tends to make errors, usually due to over-stemming (e.g., “organization” is reduced to “organ”) or to under-stemming (e.g., “create” and “creation” do not conflate to the same root).

Most of the studies so far have been involved in evaluating IR performance for the English language, while studies on the stemmer performance for less popular languages are less...
which is a derivational plural form (irregulars in Persian). There are borrowed from Arabic and some Persian words. There are some orthographic rules which show the effects of (ان) is used for humans, animals and everything that is alive. Also, “ان” (ین) is used for some words borrowed from Arabic and some Persian words. There are another kind of plural form in Persian that is called Mokassar which is a derivational plural form (irregulars in Persian). Some examples of plural form are shown in Table II. Also, there are some orthographic rules which show the effects of joining affixes to the word. For example, consider there are two parts of a word: A and B for joining as BA (Consider, Persian is written right to left). If the last letter of A and the first letter of B are “ین (ین)”, one letter “ین (ین) is added between them. Assume A is “این (اغا - mister) and B is “آن (آن), the joining result is “این (اغا-یان: men) [13].

### Table I: Some Rules for Verbs in Persian

<table>
<thead>
<tr>
<th>Rule</th>
<th>Example</th>
</tr>
</thead>
<tbody>
<tr>
<td>(present person identifier + present root + mi)</td>
<td>(I read)</td>
</tr>
<tr>
<td>(past person identifier + bud + past root)</td>
<td>(I had gone)</td>
</tr>
<tr>
<td>(present root + b)</td>
<td>(Pass)</td>
</tr>
<tr>
<td>(shod + h + past root)</td>
<td>(it was read)</td>
</tr>
</tbody>
</table>

### Table II: Some Kinds of Plural Form in Persian

<table>
<thead>
<tr>
<th>Joining</th>
<th>Result noun</th>
</tr>
</thead>
<tbody>
<tr>
<td>كنوره + ها</td>
<td>(countries)</td>
</tr>
<tr>
<td>ماه + کشور</td>
<td>(countries)</td>
</tr>
<tr>
<td>نسخته + آن</td>
<td>(countries)</td>
</tr>
<tr>
<td>تحرک + آن</td>
<td>(countries)</td>
</tr>
<tr>
<td>کشور + ب</td>
<td>(countries)</td>
</tr>
<tr>
<td>کشور + این</td>
<td>(countries)</td>
</tr>
<tr>
<td>(än + y + äghä)</td>
<td>(men)</td>
</tr>
<tr>
<td>(än + y + äghä)</td>
<td>(men)</td>
</tr>
</tbody>
</table>

### III. Persian From A Morphological Perspective

Persian is an Indo-European language, spoken and written primarily in Iran, Afghanistan, and a part of Tajikistan. It is written from right to left in the Arabic-like alphabet. In Persian, verbs involve tense, number and person. For example1, the verb “ما خوانم” (mi-xän-am: I read) is a present tense verb consisting of three morphemes. “ما” (am) is a suffix denoting first single person “من” (xän) is the present tense root of the verb and “ما” (mi) is a prefix that expresses continuity.

If a verb has any object pronoun, it can be attached to the end of the verb such as “مان گویم” (mi-xän-am : I read it) in which “مان” (sh: it) is an object pronoun. Also, negative form of verbs is produced with adding “ن” (ne) to the first of them. For example, the verb “خوانم” (ne-mi-xän-am : I don’t read) is the negative form of the verb “خوانم” (mixänam : I read). There are some certain rules to make verbs in Farsi language that some of them are shown in Table I.

There are many challengeable rules for nouns that in following, one of them is described. The plural forms of some of them are shown in Table I.

### IV. The Proposed Algorithm

A. The Proposed Method

Our Farsi stemmer is based on morphology and uses multiple phases conforming to the rules of suffix stacking. Also, it enforces a lower bound on the information a stem retains. The Farsi stemmer uses stem length to define a lower bound on information content (the minimum stem length is three). This limit is crucial when a non-suffix substring of a short word is incorrectly identified as a suffix. The Farsi stemmer identifies prefixes, and it removes prefix according to defined sequences.

The first step of the stemmer algorithm is to find a terminal substring of the input word that is in a list of common Farsi morphological prefix. Then it removes the suffix of input word. If multiple suffixes match the word, the stemmer chooses the longest suffix that would leave a stem with three or more characters. Consider the Farsi word “دستم ( thumbs)”. Both the plural suffix “آن” and the plural possessive “مان” match the end of the word. Removing “آن” leaves four letters, and removing “مان” leaves three letters. Because both leave long enough stems, the stemmer removes “مان” the longest, giving “دستم” (hand).

The suffixes are grouped as verb-suffixes, plural-noun-suffixes, possessive-noun-suffixes, other-noun-suffixes (e.g. “نام” (name)), and other-suffixes (e.g. “خان”). This grouping guides removal of prefixes from verbs and removal of multiple suffixes from a noun. If the stemmer first identifies the suffix “خان” in the word “ارزان” (“they did not
go") as a verb-suffix, it then identifies and removes the prefix "ن" to produce the stem "رفت" ("went"). Noun suffixes are stacked according to the pattern 1 (reading right-to-left):

\{Possessive\}{Plural}\{Other\} < Stem >

For example, the stemmer first finds the possessive noun suffix "نمانان" in the word "نمانانهایمان" ("our singers"), then it finds the plural noun suffix "ها" and, finally, it finds the other-noun-suffix "آن" (which signifies agency) to give the stem "نمانان" ("sing"). Hence the stemmer removes up to three suffixes from nouns.

In addition, there are some unusual cases. Usually, when the stemmer finds the suffix "نمانان" ("stan"), it removes it. However, when it is preceded by "ن", it ignores the suffix, because the Farsi suffix "نمانان" ("location of"); pronounced "stan") is often used for countries and regions, e.g. "Kurdistan". The stemmer does not remove "نمانان" because generally, the resulting conflations (e.g. Kurd = Kurdistan) are not helpful for a search engine.

Another exception is that the stemmer finds verbal suffixes "ن" and "نک" but does not remove them. That the infinitives end with "ن" or "نک". Most of the Farsi tenses are formed after removing the suffix "ن" but leaving characters "ن" or "نک". In many cases, the stemmer looks at the letter preceding a supposed suffix. Often, this pre-suffix can be used to determine whether the match is actually a suffix and, if it is, whether it ought to be removed. In such cases, if the suffix is removed, the pre-suffix remains [14]. Our first algorithms results had some problems because of the exceptions. These exceptions should be found out to improve the algorithm.

**B. Implementation**

The BNF machine is used to implement the algorithm. This implementation includes a suffix stemmer and a prefix stemmer. All suffixes will be removed during the fifteen states of the suffix stemmer. Also the prefix stemmer has two states to detect and remove the prefixes. This implementation has two final steps that will be described later. To save the detected suffixes and prefixes of each word to compare the class of suffixes or prefixes whenever it needs, two arrays are used. Suffix stemmer receives the word in reverse direction. After some proportional steps one of these following final states will be observed:

**State0:** in this state a suffix or prefix has been detected. So it will be removed and the word will be given back to the suffixstemmer or prefixstemmer as a new word.

**TABLE III: THE RESULTS OF THE ALGORITHM**

<table>
<thead>
<tr>
<th>Test number</th>
<th>Total words</th>
<th>Correct results</th>
<th>Incorrect results</th>
<th>Percentage of Correct results</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>41</td>
<td>33</td>
<td>8</td>
<td>80.50</td>
</tr>
<tr>
<td>2</td>
<td>89</td>
<td>76</td>
<td>13</td>
<td>85.40</td>
</tr>
<tr>
<td>3</td>
<td>120</td>
<td>97</td>
<td>23</td>
<td>80.84</td>
</tr>
<tr>
<td>4</td>
<td>130</td>
<td>108</td>
<td>22</td>
<td>86.07</td>
</tr>
<tr>
<td>5</td>
<td>547</td>
<td>492</td>
<td>55</td>
<td>89.94</td>
</tr>
</tbody>
</table>

Last state: the above operation is repeated until it can’t detect any suffix or prefix or the word contains less than three letters. In this case the word is returned without any removal.

Prefix stemmer acts like suffix stemmer but it doesn’t need to reverse the word. Before removing any suffix and prefix in each stage, the stemmer checks the suffixes and prefixes that were removed in previous steps and also it checks the type of the word. The current suffix or prefix will be removed, if its type is similar to previous removed suffixes and prefixes and it should be consistent within the type of the word.

**V. Evaluation and Discussion**

We select five texts with various topics on internet, and test them with our algorithm. The results are shown in Table III and Fig. 1.

As it can be seen, by increasing the quantity of tested words, the average percentage of errors is decreased. It should be noted that many of these errors are due to non-verb words that are structurally similar to verbs or because of the verbs that their stems are less than three letters. This algorithm is changed as following due to mentioned problems.

**A. Improvements on the Algorithm**

There are some words that are structurally similar to other words. These words should not be used by prefix stemmer and suffix stemmer. For example the first letter of non-verb word "پربرمان" is "پ" which is same as the prefix of imperative verbs in Farsi. But the letter "پ" should not be removed. Or the word "پیمکت" starts with "پ" which is similar to negative verbs and it ends with "ت" that is same as possessive pronoun. But these letters should not be removed as prefix and suffix.

Also there are some plural words in Farsi, named Mokassar which there are no certain rules to make them. Current rules couldn’t be used to find these words stem.

Furthermore this algorithm has a restriction which the resulted stem should have three or more letters. But there are some words that their stem's length is less than three. For example for the verb "پیریشکیم" the algorithm removes the term "پیری" as prefix. Then it detects "پ" as the longest suffix, but if the algorithm removes "پ" the remind part will have only two letters. So it removes just "پ" and returns "پیریشکی" as the stem, while the correct stem is "پیریشکیه".

So a data base is used to save these words stem and the algorithm is improved by considering these exceptions.

**B. Improvements on the Implementation**

A data base is used to improve the stemmer. In this database, non-verb words which start with a term that is similar to a verb maker prefix or words which end with a term that is similar to a suffix are saved. But if after removing...
these terms, the remained part of word had less than three letters, these words should not be saved in database.

For example the word “"میوه"” starts with "می" which is similar to verb maker prefix "می" in Farsi. But it’s not a prefix. If the part "می" is removed, the remained part "یوه" will have only two letters. So this word should not be saved in database. Or the word ""نیمه"" starts with "نی" which is similar to negative verb’s structure. If the term "نی" is removed, the remained part will have three letters. Therefore the word ""نیمه"" should be saved in database. Also some plural words named mokassar and their singulars are saved in database. At the start of algorithm, the word should be searched in database. If it is found, its stem will be returned. Otherwise it will be used by algorithm’s functions to remove suffixes and prefixes.

Furthermore, some words which their stems have less than three letters are saved in database. If after removing the suffixes or prefixes the stemmer confronts a stem with less than three letters, at first it will search the database. If the stem is found in the data base, it will be returned. But if it isn’t found, the stemmer doesn’t remove the suffix or prefix.

C. Comparative Evaluation

To evaluate these two algorithms, four different texts were selected from internet and were tested by these algorithms. The results are shown in Table IV and Fig. 2.

<table>
<thead>
<tr>
<th>Test number</th>
<th>Percentage of Correct results (first algorithm)</th>
<th>Percentage of Correct results (Improved algorithm)</th>
<th>Time (first algorithm)</th>
<th>Time (Improved algorithm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>84.25</td>
<td>94.76</td>
<td>5 sec</td>
<td>5 sec</td>
</tr>
<tr>
<td>2</td>
<td>81.65</td>
<td>97.95</td>
<td>5 sec</td>
<td>6 sec</td>
</tr>
<tr>
<td>3</td>
<td>88.68</td>
<td>97.86</td>
<td>17 sec</td>
<td>18 sec</td>
</tr>
<tr>
<td>4</td>
<td>88.14</td>
<td>96.78</td>
<td>25 sec</td>
<td>25 sec</td>
</tr>
</tbody>
</table>

![Fig. 2. Compare of two algorithms](image)

As mentioned before, to improve our algorithm a data base is used that some exceptions are saved in it. This evaluations show that the percentage of correct results is increased while the speed of algorithm doesn't change.

VI. CONCLUSION AND REMARKS

In this paper the stemming is described. Then the applications of stemming and different types of stemming algorithms were explained. At first an algorithm is implemented based on morphology and its problems were described. Afterward a modified algorithm was presented to improve the results. In the proposed method a database is used which contains some exceptions and based on morphology. Morphology is used to find the stem of the words. The stemmer was improved by saving the words that are similar to other words structure, and also some exceptional plural words named Mokassar and some stems that have less than three letters in a database. The number of these words is low in compare with the number of all Farsi words. But this algorithm is depended on database and in some cases the result is wrong because the stemmer can't detect the type of the words. This problem will be solved by finding out the type of the words according to the structure of the sentences.

REFERENCES

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