# A Hybrid Stemmer of Punjabi Shahmukhi Script 

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#### Abstract

Summary Stemming is a heuristic process to chop off end part of words and sometimes adding additional letters at the end of words to get the basic meaningful forms of surface words. The basic goal of stemming is to reduce inflectional forms of words to root words using multiple techniques. In this paper, hybrid approaches are used for stemming Punjabi words. There has not been any stemmer reported for Punjabi شاه مكهى (Shahmukhi) script. We used database lookup approach and rule based stemming for Punjabi Stemmer. Our dataset consists of 2.5 million tokens which were divided into three parts of 1500000, 500000 and 500000 tokens and used for training, development and testing purpose respectively. We got $86.01 \%$ accuracy while tested our stemmer over above specified dataset by using 63 rules.


## Keywords:

Rule based stemmer, morphology, lookup approach, root words, hybrid stemmer, affixes and normalization.

## 1. Introduction

Stemming is a process to strip off root words and connected sub words from surface words. Root words are called stems and connected sub words are known as affixes. For example, in English "making and makes" both will be stemmed to root word "make". Here stem is "make" and affixes are "ing" and "s", respectively. Similarly, in Punjabi, word كرُّيا Kuryan (Girls) would be stemmed to كزُى kuri (Girl). More examples are given in Table 1.

Table 1: Examples of Punjabi شاه مكهى (Shahmukhi) Stemmer

| Sr. | Inflected words | Stem / Root |
| :---: | :---: | :---: |
| 1 | كزّيان Kuryan (Girls) | $\begin{gathered} \hline \text { كزُى Kuri } \\ \text { (Girl) } \\ \hline \end{gathered}$ |
| 2 | $<$ من Munday (Boys) | مندًا Munda (Boy) |
| 3 | منتان Mintaan (Appeals) | منت Minatt (Appeal) |
| 4 | كتابان Kitaaban (Books) | $\begin{gathered} \text { كتّاب Kitaab } \\ \text { (Book) } \end{gathered}$ |

Punjabi is a language, which is widely spoken in Pakistan and India. Almost 100 million people have Punjabi as their native language in Pakistan, India, Canada and other European countries [16]. It has two types of scripts, one is گگرمكهى(Gurumukhi) which is usually written in Hindi script and is spoken in Indian Punjab. Other script is شاه مكهى (Shahmukhi) script which is written in Arabic script
and follows Right to Left (RTL) language styles. Sufficient work has been done on گُرمكهى(Gurumukhi) regarding information retrieval and text processing.
ششاه مكهى (Shahmukhi) is usually used in Pakistani Punjab for communication. Existing work using this script is deficient and disappointing.
Urdu also uses an RTL style and has the same script as شاه مكهى (Shahmukhi), which supported us to make rules for our stemmer by using previous work done in Urdu like in [7].An important thing is that as Punjabi is morphologically rich language, due to which we have observed multiple scenarios in our work where stemming leads us towards wrong decisions like Over Stemming (stemming of a part which should not be stemmed) and Under Stemming (stemming of a part which should be stemmed). To handle these kinds of issues, we used additional techniques after stemming to make sure that the under and the over stemming are done properly. Table 2 shows some under and over stemming examples as follows. In this paper, the Punjabi stemmer is proposed for شاه مكهى (Shahmukhi) script by using the hybrid approaches.

Table 2: Examples of false stemming

| Sr. | Word | Stem/ <br> Root | Stemming type | Correct Stem |
| :---: | :---: | :---: | :---: | :---: |
| 1 | كاركن kaar-kun (Worker) | $\begin{gathered} \text { Kaark } \\ \text { Kارك } \end{gathered}$ | Over stemming | كاركن <br> Kaar-kun |
| 2 | بِينّوْوان <br> Pendu-waan <br> (villegers) | Painduo | Under stemming | $\begin{gathered} \text { يיنظّ } \\ \text { Paindu } \end{gathered}$ |
| 3 | غدار ghaddar (Traitor) | $\begin{gathered} \dot{\varepsilon} \\ \text { Gayin } \end{gathered}$ | Over stemming | غدار Gaddaar |

## 2. Background \& Related Work

Lovins [1] discussed first about rule based stemmer for English. In this stemmer, almost 260 rules were used for stemming English words. One of the famous work done is by Porter [2] which is known as the rule based stemmer for English. He actually enhanced and simplified the work done in [1] and defined just 60 rules which are now famous due to versatility and usage. He proposed a rule based stemmer, in which series of rules were defined and words were categorized for stemming into classes and

[^0]each class of words had specific rules. Porter's work was the best work in English stemming and it helped to draw conclusions about stemming on rule based approaches as it used just 5 steps for complete stemming.
A considerable work is also done on languages of Asian region. There are different stemmers for the RTL languages like Urdu, Arabic and Persian. As Punjabi is also RTL so the work done in above Asian region languages is helpful for Punjabi شاه مكهى (Shahmukhi) Script stemmer. Urdu is quite similar to Punjabi and both use the Arabic script for writing. In [3] Husain used unsupervised approach for the Urdu stemming. He has done his work using N -grams and using two different techniques for stemming purpose. His algorithm is simply a 5-step algorithm, which first uses N -gram over Urdu corpus and splits suffix and stems. Then sort stems and suffix based on frequency and makes rules based on above 3 steps. One important point is that all those suffixes, which have count less than specified threshold are discarded and not included while making rules. Then two approaches were used, either to go for frequency-based approach or for length-based stripping off. Frequency based approach produced accuracy about $84.27 \%$ and the length-based approach gave $79.63 \%$ maximum accuracy for Urdu.
Gupta et al., suggested rule based stemmer for Urdu. They achieved accuracy up to $86.5 \%$ using different forms and variations of rules. Interesting thing in their suggested rules was that they categorized words into main categories like verb, adjective, noun etc., and provided surprising results based on category specified rules. Further, there were repetitive experiments performed to make sure that the extracted suffixes and stems were meaningful and logical. [4]
The work in [5] was a refined and updated version of [6]. Here newspaper corpus was used for stemming purpose. Database of suffixes was used to improve results and suffixes were categorized into 75 possible items. An efficient algorithm was provided which not only stemmed properly but also resulted $90 \%$ precision. This idea enhanced the accuracy, as this was the combination of algorithmic approach, use of a database and the rules for stemming.
Akram et al. [7] proposed an Assas-Band stemmer, which not only stem words but also improve the efficiency by maintaining proper classes and categories. For example, it maintained the distinction between feminine and masculine while stemming, which was a plus point for this rule-based stemmer. The technique used was simple, as a word is composed of prefix, stem and postfix so in first attempt, prefix was removed then postfix was removed and finally, the stem was processed further for exception handling. It also included rules list, which was used to remove, prefixes and postfixes from the words to get stems. After applying the techniques discussed, additional
changes were also required in special cases to add extra characters to make the stems proper and meaningful words. In this phase, extra characters were added to enhance the efficiency of the overall process. In the last stage, manual corrections were applied to check whether proper classes' distinction has been maintained like masculine and feminine class or lookup approach has been used to verify any missing entity in the Prefix and the Postfix list, which were used for the initial processing. Overall maximum accuracy achieved was 91.18 percent.
A hybrid approach was used by Thapar in [8] for the Punjabi Stemmer of گرمكهى(Gurumukhi) script which is written in Hindi. Whenever a word was provided as input, it was searched in a database of root words, which had been already created by collecting the data. If a word found in the database, then it was not stemmed and displayed as a stem. If it was not found in the database, then proper stemming techniques were applied to find out root word. Similar kind of work has been done by Joshi \& Garg in [9]. This paper included 3 techniques for stemming Punjabi words of گرمكهى (Gurumukhi) script. These 3 techniques included brute force approach, rule based approach and synset approach. Words were searched in tables if they found then it gave 100 percent accuracy. If not, then proper stemming was performed and after that over-stemming and under-stemming were properly handled in this stemmer. Further, to handle such suffixes, reduction or substitution techniques were used, which gave root words and then these words were verified by the lookup approach finally.
K Raiz in [10] highlighted challenges faced in Urdu stemming. Urdu being a rich language uses words from many other languages like Persian, Arabic, etc. Urdu has the Arabic and Persian orthography and also shares the grammar rules. Raiz explored the rule-based approach further with the following attributes like nouns, adjectives, pronouns, verbs, numbers, date \& time, Persian loan words, Arabic loan words, gender \& number agreement, etc. Other challenges that were described in his work by K Raiz, are the engineering issues. Out of 569 words, 211 were stemmed and out of these 211 words, 32 words were stemmed incorrectly. V. Gupta \& G. S. Lehal worked on keyword extraction for Punjabi in [6]. According to them, this can be achieved in different phases like removing the stop words, identification of Punjabi nouns and noun stemming, calculating term frequency and inverse sentence frequency etc. Bundle of experiments were performed over 50 Punjabi documents for text extraction. The result of those experiments found precision on $80.4 \%$, recall on $90.6 \%$ and F-Score on $85.2 \%$. There was certain percentage of errors like $14.8 \%$ which was due to dictionary mistakes, absence of certain Punjabi nouns, syntax mistake in input text and violation of rules for certain noun stemming.

Stemmer 1 for Quran (the book of God) used two type of stemming algorithms: prefix stemming and suffix stemming. All the stop words were excluded from the stemming process and this was achieved by using Quranic lexicon prepared by Fuad 'Abd Al Baqi [12]. The accuracy of stemmer was $99.6 \%$ for the prefix stemming and $97 \%$ for the suffix approach. Inaccuracies detected in the results were because of incorrect transliteration of lexical items [13].
Brute force approach as termed as exhaustive search was used to build the Punjabi stemmer [14]. In this approach a lookup table was used which possessed the relationship between the root word and the inflected word. If the word was not found in lookup table, then suffix stemming or suffix-striping algorithm came into the role. Accuracy of this stemmer was highly dependent on the number of words presented in the look up table. Larger the lookup table, better were the results.
V. Gupta \& G. S. Lehal proposed the stemming for the Punjabi nouns and proper names. According to them, if stem words are present in the dictionary then it is noun else it is a proper name. Experiments were performed on the corpus of Punjabi and found $87.37 \%$ efficiency. They used rule base approach and created different orthographic rules but overall, error percentages were $9.78 \%$ due to rule violation, $2.4 \%$ due to dictionary mistakes and $0.45 \%$ due to spelling mistakes. [15]
Punjabi is similar to Urdu and in our stemmer, we used hybrid approaches which are discussed in the next section.

## 3. Proposed stemmer

We have proposed a hybrid stemmer for Punjabi شاه مكهى (Shahmukhi) script. We have used a rule based and a lookup based approach for this stemmer.
The workflow is described in following algorithm and the basic flow of the work is given in the form of flowchart in Fig 1.

```
ALGORITHM 1. Hybrid Punjabi شاه مكهی (Shahmukhi) Stemmer
Input: Set of Punjabi words.
Output: Properly stemmed roots.
    1. Read input
    2. Tokenize on space basis
    3. Pre-Process/ Normalize (this includes removal of
        special characters and organizing the words properly)
    4. For each word \(W_{i}\) in Tokens
            If \(W_{i}\) belongs to Exception List or Lookup Table
                    Output stem against \(W_{i}\) and continue to step4;
            If wordLength \(\left(W_{i}\right)<=3\)
                    Apply rule 4 given in Table 11 and move to
                    step 4
```

[^1]
## Else

Remove prefix/postfix;
Apply rules on $W_{i}$, output processed stem; If output word $W_{o}$ belongs to dictionary, then add it to Lookup Table against $W_{i}$; Move to step 4;


Fig 1: Flow chart of Punjabi Stemmer.

### 3.1 Details of Algorithm

The details of algorithm include different modules of our stemmer. All of these modules are sequential. Each module is described precisely as follows.

### 3.1.1 Tokenization

Input words are split based on space as a delimiter. After this phase, we have an array of words known as tokens, which is used further for stemming process.

### 3.1.2 Normalization

In normalization phase special characters @, \#, *, \&, etc., are removed so that better results could be achieved just in handling Punjabi text 2 and characters. In this phase, English and other languages text and words are removed.

### 3.1.3 Iterative Approach

Iterative approach is used to process the whole input dataset. All words are processed sequentially and during each iteration, hybrid3 approaches are used to get stem of the under process word.

[^2]
### 3.1.4 Length Based Decision

If length of a word is less than 4 characters and it is already in root form, then there is no need to process that word. Examples are provided in Table 3.

| Sr. | Word | Stem / Root |
| :---: | :---: | :---: |
| 1 | مجه Majj (Buffalo) | ${ }_{6}$ مMajj |
| 2 | مان Maan (Mother) | ماMaan |
| 3 | جٌ Chatt (Roof) | -جهٌ |

### 3.1.5 Exception List

Exception List includes all those words on which we can't apply any rule. These words should be included in output file as it is. These all words are maintained in a database list known as Exception List. All kinds of nouns and proper nouns can also be the part of this Exception List to produce accurate results. Table 4 and Appendix B show examples of such words.

| Sr. | Word | Stem / Root |
| :---: | :---: | :---: |
| 1 | بآتّهى Haathi (Elephant) | -بآتهى Haathi |
| 2 | ¢اككستانPakistan | ¢اككستانPakistan |
| 3 | كهُّارى Khdaari (Player) |  |

### 3.1.6 Prefix/Postfix Rule

We have two lists, which have all prefixes and postfixes. Word is checked against any of the prefix or postfix. If any prefix or postfix is found, it is removed. Table 5 and 6 show the lists of prefixes and postfixes, while the Table 7 presents the example of prefix and postfix rule.

Table 5: List of Prefixes.

| ب | نو | بد | ما | تو | ك |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\begin{aligned} & \text { Beh+A } \\ & \text { laf } \end{aligned}$ | Noon+Waw | Beh+ <br> Daal | Meem+Al af | Teh+ Waw | Kaaf+ <br> Meem |
| C | بالا | L | مبا | V | ان |
| Beh+B | Beh+Alaf+ | Noon+ | Meem+He | Laam+ | Alaf+ |
| ri-Yeh | Laam+Alaf | Alaf | h+Alaf | Alaf | Noon |


| باز | دار | فروش | كابى |
| :---: | :---: | :---: | :---: |
| $\begin{aligned} & \text { Beh+Al } \\ & \text { af+Zeh } \end{aligned}$ | $\begin{aligned} & \text { Daal+Alaf } \\ & + \text { Reh } \end{aligned}$ | Feh+Reh+Waw +Sheen | $\begin{aligned} & \text { Gaaf+Alaf+Heh+Choti } \\ & \text {-Yeh } \end{aligned}$ |
| كار | ناك | خور | بين |
| Gaaf+A | Noon+Alaf | Kheh+Waw+R | Beh +Choti- |
| laf+Reh | +Kaaf | eh | Yeh+Noon |

Table 7: Prefix/Postfix based decisions.

| Sr. | Word | Stem/ root | Rule Type |
| :--- | :--- | :--- | :--- |
| 1 | بااصول Baa-Usool (Disciplined) | دغاز Dagha Baz (Treacherous) | باز Unool |
| 2 | دغا Drefix |  |  |

### 3.1.7 Lookup Table

A majority of words are stored in Lookup Table, a database table. These words are stored along with their stems. A word is first searched in Lookup Table. If a word belongs to the table, then its stem is returned and this shows 100 percent accuracy for that specific word. It is a little bit different from the Exception List. In Exception List words are added before processing and in dictionary words are added along their stems after complete processing. This Lookup Table along with the Exception List is verified manually and it enhances our accuracy and efficiency.

### 3.1.8 Rules Based Approach

In this phase, it is considered that a given word is totally unseen, as it does not belong to the Lookup Table or the Exception List. To handle such words, we have defined a set of 63 rules, which stems these words properly.

### 3.2 Algorithmic Rule Based Approach

Few important rules are mentioned here for all words of length $>3$, which are also not in the Lookup Table or the Exception List.

### 3.2.1 Rule 01

If a word ends with ال (Alaf + Noon Gunna) then remove ال from the end of the word. Examples are given in Table 8.

| Sr. | Word | Stem/ root |
| :---: | :---: | :---: |
| 1 | ونگان Wangaan (Bangles) | ونگ, Wangg |
| 2 | قپتان chatatan <br> (Roofs) | Chatt |

### 3.2.2 Rule 02

If a word ends with $\angle$ (Bari yeh) then remove $\angle$ from the end of the word and insert I (Alaf). Table 9 shows a few relevant examples of this rule.

| Sr. | Word | Stem/ root |
| :---: | :---: | :---: |
| 1 | $\begin{aligned} & \hline \text { كان Gaany } \\ & \text { (Songs) } \end{aligned}$ | كانا <br> Gaana |
| 2 | ج جوو Choowy <br> (Rats) | جووا Choowa |

### 3.2.3 Rule 03

If a word ends with ين (Choti yeh+Noon) then remove ين from the end of the word. Examples are given in Table 10.

| Table 10: An example set for Rule 03. |  |  |
| :---: | :--- | :--- |
| Sr. Word <br> 1 ششوقين Shoqeen (Fond of) <br> 2 Rangeen (Colorful) | Stem/ root |  |

### 3.2.4 Rule 04

If a word length is between 3 and 5 and the word ends with g (Waw) then remove 9 from the end of the word. Examples are given in the following Table 11. Based on the variation and different variant forms of the words, there is a variety of words in Punjabi. If words length is less than 4 then there is more than 98 percent chance that the word itself is root.

| Sr. | Word | Stem/ root |
| :---: | :---: | :---: |
| 1 | ورك Kro (Do this) | SJkar |
| 2 | وهكى Waikho (Look) | هكى Waikh |
| 3 | ¢ֻج Pajjou (Run) | ج Pajj |

All our rules cannot be presented here. Few important ones are discussed and the rest are depicted in appendices.

## 4. Experimental Evaluation

For our Punjabi stemmer, we collected dataset from online resource [11] of Punjabi. In this section, experimental details over specified dataset of Punjabi are prescribed.

### 4.1 Experimental Dataset

We have collected dataset of 2.5 million words which has 85152 unique words from online resources [11]. Our dataset is mixture of literature, politics, and science and poetry words. We divided data into three parts i.e. training data C1, development data C2 and testing data C3. Using C1 we have developed initial list of rules and then using C2 we improved our rules to get better results. Table 12 shows the number of tokens and unique words in each part.
Table 12: An overview of corpus

| Corpus Number | Total words | Unique words |
| :---: | :---: | :---: |
| C1 | 1500000 | 35168 |
| C2 | 500000 | 25073 |
| C3 | 500000 | 24911 |
| Total | 2500000 | 85152 |

### 4.2 Minimum Word Length Rule

During experiments, we have observed that 98 percent words of length less than 4 need no stemming and these are already in base or root form. Few examples are given in Table 13.

Table 13: Example set for MinWordLength rule

| Sr. | Word | Root / Stem |
| :---: | :---: | :---: |
| 1 | كر Karr (Do) | كا Karr |
| 2 | $ج ا$ Jaa (Go) | جا Jaa |

$$
\begin{array}{|l|l|l|}
\hline 3 & \text { بول Bol (Speak) } & \text { بول Bol } \\
\hline
\end{array}
$$

### 4.3 Lookup Table Updating Rule

In our first experiment, we used the rule-based approach over our corpus C 1 . The results are not good and there is a verity of under and over stemming. An example set is given in Table 2. All those words, which are not stemmed properly then added into the Lookup Table with their correct stem and updated Exception List.

### 4.4 Rules Revision Policy.

While working over 1.5 million words of Punjabi corpus, we have observed that few rules are giving inaccurate results and instead of enhancing accuracy, these rules are decreasing efficiency of our stemmer. Updated rules are given in Appendix A.

### 4.5 Accuracy Calculations.

Accuracy of our stemmer is calculated through equation 1. Accuracy (\%) $=\frac{\text { Correctly stemmed words }}{\text { Total tested words }} * 100$ (1)
We have used C2 for Phase 01, Phase 02 and Phase 03 and C3 have been used for Phase 04 experimentation.

### 4.6 Stemming Process

Stemming process is divided into the following phases.

### 4.6.1 Phase 01

In Punjabi Stemmer, the rule-based approach has been used in Phase 01 of accuracy calculations over the development data C2. An applied algorithm given in section 3 has been tested on 25073 words and has got 8956 words stemmed properly with accuracy $35.71 \%$. Details are provided in Table 14.

### 4.6.2 Phase 02

In second phase, we have manually put the correctly stemmed words into the Lookup Table and updated our rules based on the remaining un-stemmed words. We have learnt from experiments in this phase that few words are better not to stem up to the last level, for these kinds of words, better approach is to stem up to a reasonable extent. For example, بدمعاش (Badd-Muaash) would be stemmed to معاش(Muaash) instead of شاعمدب (Badd-Muaash). In fact, few words should not be stemmed or stemmed to a reasonable extent. Examples are given in Appendix B.
For phase 02, the dataset C2 is used and we have applied both the approaches: rule based approach and the Lookup Table approach. For this phase, we have used 25073 unique words and we have got 17804 words correctly
stemmed with accuracy of $71.01 \%$. Results are listed in Table 14.

### 4.6.3 Phase 03

In this phase, rules are modified and the Exception List is included and we put all those words in EXCEPTION, which occur in our corpus with a huge frequency and follow no rules. We populated our Lookup Table by correcting the under and over stemmed words manually. For phase 03, we have 26494 words in our Lookup Table help in improving the accuracy, which is clear in Table 14. Some of the rules are defined in Appendix A.

### 4.6.4 Phase 04

In this phase, dataset C3 is used. We have modified rules and updated Lookup Table with 30494 words. All these words are properly stemmed and we verified accuracy of our testing data through manual comparison.
We used 0.5 million words data set having 24911 unique words. From which our algorithm correctly stems 21426 words. Results are mentioned in Table 14 and Fig 2.

Table 14: Phase wise results.

| Phase <br> No. | Total <br> words | Unique <br> words | Correct <br> stemmed | Accuracy <br> $\%$ |
| :---: | :---: | :---: | :---: | :---: |
| 01 | 500000 | 25073 | 8956 | 35.71 |
| 02 | 500000 | 25073 | 17804 | 71.01 |
| 03 | 500000 | 25073 | 20517 | 81.82 |
| 04 | 500000 | 24911 | 21426 | 86.01 |

## 5. Results \& Discussions

We used the Lookup Table, Exception List and the repeatedly modified rules. During phase 02 and 03 , we modified, eliminated and added new rules to improve the accuracy of the proposed stemmer. In the end, Lookup Table became so large to handle almost majority of unseen words. Currently, the Lookup Table consists of 30494 unique words, which can be increased to achieve more accuracy. Proposed stemmer produced $86.01 \%$ accuracy using both the rules and the Lookup Table. Without using Lookup Table, proposed stemmer produced 35.71\% accuracy in phase 01 and $64.90 \%$ in phase 04.


Phase Wise Accuracies of Stemmer

Fig 2: Stemming accuracy in all phases.

## 6. Limitations

Proposed stemmer uses rule based and lookup based approaches for stemming Punjabi words. As Urdu and Punjabi has a lot of common vocabulary, so, the proposed stemmer by default handles the Urdu data up to some extent. Proposed stemmer is efficient when it uses both rules and the Lookup Table. Rules are defined, which covers major part of Punjabi data but not all words.

## 7. Conclusion \& Future Work

The proposed Punjabi stemmer for شاه مكهى (Shahmukhi) script first uses the length based decision to output the proper stem. After taking this length-based decision, for words greater than 3 characters' length first it uses the lookup approach, which includes lookup from the Lookup Table and the Exception List. This stemmer is first such stemmer for شاه مكهى (Shahmukhi) script and gives us promising results with 86.01 percent accuracy by using the prefix, postfix and infix removal methods. This accuracy can be improved by improving the Lookup Table. In future, the rules should be increased and the Lookup Table should be enlarged, this will give us more accuracy for our stemmer. Furthermore, statistical approaches can also be used to decide the best root for a given word.

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Appendix A: Few Rules for Punjabi Stemmer

| Rule Description | Examples. Word Stem |  |
| :---: | :---: | :---: |
| If a word ends with نا (Noon+Alaf) then remove ul from end of the word | اناج Jana | اج Jaa |
| If a word ends with (Kaaf+Alaf+Reh) then remove راك from end of the word | راك حالص Slah Kaar | حالص <br> Slaah |
| If a word ends with ulوا (Alaf+ Waw+Alaf+Noon -gunnah) then remove ulo from end. | ט\|واهكس Sikhawan | اهكس <br> Sikhaa |
| If a word ends with (Gaaf+Alaf+Reh), remove راگ from end of the word. | راگددم <br> Madad Gaar | ددم <br> Madad |
| كیان If a word ends with (Noon + Alaf + Kaaf), remove كان from end of the word. | كاندرد <br> Dard Naak | $\begin{aligned} & \text { درد } \\ & \text { Dard } \end{aligned}$ |
| If a word ends with روخ (Kheh + Waw + Reh), remove روخ from end of the word. | روخ مردآ Adam Khour | مدآ Adam |
| If a word ends with $\mathbf{S}^{*} \leftharpoonup$ (Choti-yeh+Any character +Bri-yeh) then remove $\_$from end of the | <çسهگ Ghseety | -عسهگ Ghseet |


| word. |  |  |
| :---: | :---: | :---: |
| If a word has lenth >=6 and word ends with (Daal + Choti yeh + Alaf+Noon gunnah) then remove slu from end of the word. | טایدكشش Lashkdyaan | كشل Lashak |
| If a word starts with شخ (Kheh + Sheen) then remove شخخ from start of the word. | طخ شخ <br> Khush khatt | شخ Khush |
| If a word starts with $n$ (Heh + Meem) then remove $\sim$ م from start of the word. | رمع م~ <br> Hamm Umar | ر ر Umar |
| If a word starts with هدا (Alaf +Daal+ Heh) then remove هدا from start of the word. | ایوم هدا Adh moya | احوم <br> Moya |
| ان If a word starts with (Noon + Alaf) then remove نا from start of the word. | دارم ان <br> Na muraad | دارم Muraad |
| If a word ends with نو (Waw + Noon) then remove نو from end of the word | نواكمج٪ Chamkawann | اكمج Chamka |
| If a word ends with ن (Noon) then remove ن from end of the word. | نرك Karan | رك Karr |
| If a word ends with راد (Daal + Alaf + Rey) then remove روخ from end of the word. | راد نامى\| Emaarn Daar | نامی\| Emaan |
| If a word ends with $u$ (Waw + Noon gunnah) then remove 90 from end of the word. | نول\|ن <br> Naaloon | لان <br> Naal |
| ن If a word ends with (Laam + Noon) then remove $\cup$ from end of the word. | نلوب <br> Bolann | لوب Bol |
| If a word has length >=5 ends with 1 (Alaf+Waw+Bri-yeh) then remove l ¢ from end of the word. | <ورارّل Laraway | ر Larr |
| If a word ends with luد」 <br> (Alaf+Noon+Daal+Bri- | ^ <br> Mangwan-day | اوگنم <br> Mangwa |


| yeh) then remove ن $\_$from end of the word. |  |  |
| :---: | :---: | :---: |
| If a word has length $>5$ and ends with ulوg (Waw+Waw+Alaf+Noon gunnah + Teh) then remove le from end of the word. | טاووجگنج <br> Jangjouwaan | $\begin{aligned} & \hline \text { وجن-Jang-Ju } \\ & \hline \text { Jan } \end{aligned}$ |

Appendix B: Exceptions

| Rule Description | Example | Exception word |
| :---: | :---: | :---: |
| If a word ends with $ى^{*} \leftharpoonup$ (Choti-yeh+Any character + Bri-yeh) then remove $\angle$ from end of the word. | $\stackrel{\text { Ghseety }}{\text { گ. }}$ | $\xlongequal[\text { Saleeqy }]{\text { سليق }}$ |
| If a word ends with اu (Alaf+Noon gunna) then remove lum end of the word. | عورتان Aourt-aan | دعاوان <br> Duawaan |
| If a word length>=5 and word ends with $\sim_{\sim}^{\text {l }}$ (Alaf + Noon+Bri-yeh) then remove $\underset{\sim}{-}$ from end of the word. |  | $\overbrace{\text { Kamaany }}^{\text {كمان }}$ |
| If a word length $>=6$ and word ends with اندا <br> (Alaf + Noon + Daal + Bri-yeh) then remove $<$-ifrom end of the word. | منگواد <br> Mangwa day | بپ <br> Parandy |
| If a word has length>3 and word ends with $\underset{\sim}{\sim}$ (Hamza Choti yeh +Bri yeh) then remove $\simeq$ from end of the word | $\underset{\text { Sikhaaye }}{\substack{\text { سيكها } \\ \hline}}$ | Áf |
| If a word has length >=5 ends with $<$ او (Alaf+Waw+Bri-yeh) then remove $\angle$ glfrom end of the word. | لزُاو <br> Laraway | جاو Jaawy |
| If a word ends with ين (Chotiyeh+Noon) then remove ين from end of the word. | شوقين Shoqeen | أينين Aaein |
| If a word starts with بد (Beh+Daal) then remove بد from start of the word. | بدصورت <br> Badd-Sortt | بدتر <br> Badd-Tarr |
| If a word starts with مبا (Meem+Heh+Alaf) then remove مrfrom start of the word. | مهاراجا MahaRaja | مباتّما <br> Maha- <br> Tama |


[^0]:    Manuscript received August 5, 2017
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[^1]:    ${ }^{1}$ It was implemented using Delphi that is an objectoriented language and this stemmer was developed for windows.

[^2]:    ${ }^{2}$ Punjabi is a language composed of words of Arabic Script. It is Right to Left (RTL) language and it contains mixture of Urdu, Arabic and Punjabi text words.
    ${ }^{3}$ Hybrid approach includes lookup base approach followed by a rule base approach.

