

Urdu Spell Checking: Reverse Edit Distance Approach

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Abstract

Spelling of words of a language are standardized by language authorities or consortiums and available in dictionaries or lexicons. For instance, “produkt” does not belong to English dictionary. Similarly “درمیان” is a correctly spelled word, while “درمیانر” is a non-word in Urdu. Electronic representation of text is commonly used in today’s computing environment. The rich resourced languages like English have many applications with added tools. On the other hand, application development is in its infancy for less resourced language like Urdu. Spelling plays a vital role while humans write text electronically in computers. It is oblivious that terabytes of text is added in form of corpus or otherwise that is required to be spell checked, which is practically impossible to be done manually. In this work, various techniques for spellchecking have been studied and analyzed. All of them separately or a combination thereof can be used for the process of spell checking. Edit Distance technique has been widely used in spell checkers of various language, and a variation of this technique i.e. Reverse Edit Distance technique selected for suggesting correct words for nonwords. For Urdu, candidates are found by making $86n+41$ comparisons for an ‘n character’ length Urdu word.

1 Introduction

Usage of computers became an essential component in human lives. In today’s computing environment, text processors, search engines, short messaging services, chatting applications, and many more are widely used. Google auto complete feature starts giving options; For instance when “morp” is typed options like: morphine, morphology, morphine drug, and morphological choices appear. On the other hand the user does

not know the exact spelling, Google will make a search based on closed matched words and will display the message as well. All this happens in a fraction of second. This reveals that Google and other applications are using spell error detections and correction algorithms which are implemented in applications for user facilitation. Terabytes of text added to the internet resource daily is required to be spellchecked. A single book requires multilevel readers for spell checking and correction, if performed manually. Thus automated spell checkers and correctors are required. Spell checking is process of matching a given word with alphabetically ordered words in a dictionary or lexicon. For instance, in Urdu, a given word شاپین is compared with words in Urdu lexicon, if we get a match then the word is considered to be correctly spelled. On the other hand if we compare شاحین with words in lexicon, if we do not get a match then the word is considered to be misspelled. However it is worth mentioning here that the automated spell checking process is reasonably complex in case of Urdu as compare to English. The complex process of spell checking in Urdu is due to its morphologically richness, word space problem, and scarcely available electronic resources.

The spell checking process is a pre-requisite for language processing systems, e.g. Grammar checker, Part of speech (PoS) tagging, Information extraction, Machine translation, etc. The input to above mentioned language process systems must be correctly spelled, and the text must be passed through a dependent spell checker.

Section 2 narrates about the challenges exists about spell checking in Urdu, Linguist study is narrated in section 3, section 4 discusses proposed technique for Urdu spell checking, the review of review of reverse edit distance algorithms and discussion are presented in section 5, and section 6 is for conclusion.

2 Challenges of Urdu Spell Checking

Tokenization in Urdu, Diction problem, Loan Words, Morphological nature, Grammatical words and Initial letter Capitalization are some of the challenges that makes the Urdu Spell checking complex. Tokenization is process of to break text at word level.

2.1 Word Separator – space character

In English text, words are separated by spaces characters, as tokens (Sara Stymne 2011). To separate words in Urdu, spaces are not used after every word as in case of English language. The natives of Urdu language separate the words from each other by cognitive knowledge of the language while reading or writing text. For instance, in the following example same sentence is written in three variation of space usage: Space is only used after a word that ends with joiner letter. ہم بازار میں کھیل رہے تھے۔ [We were playing in the bazaar]. Space is not used at all after any of the word that ends with joiner letter. ہم بازار میں کھیل رہے تھے۔ Space is used after each word. ہم بازار میں کھیل رہے تھے۔

2.2 Morphological Nature

In contrast to languages like English, Urdu language bends toward agglutinative languages due to its complex morphological nature. The languages like Turkish and Finnish are agglutinative languages as multiple words generates from a single word by affixation, derivational, and inflectional suffixes. From the Turkish word “uggar” (means civilized) a word “uggarlastiramayabileceklerimizdenmissinizcesine” is derived (Kemal Oflazer and Cemaleddin 1994). “Heater”, “heated”, “heats” are the word forms that are inflected from the root word “heat”. Urdu is morphological rich language and multiple words are inflected from a root word. The following are few inflected words that are inflected from the root word of Urdu بول (speak).

تو بول
تم بولو
تم بولنا
آپ بولیں
آپ بولیں
آپ بولتے
آپ بولتے
آپ بولو

2.3 Diction Problem

Diction problem is defined as using choice of words from a set of word which has same meanings. The Urdu language is considered to be

computationally complex due to its diction problem as well. Example words in Urdu that are different in spelling but having same meanings are: (تکیا، تکیہ)

2.4 Loan Words

The native speakers of the Urdu language take advantage of **loan words** from other languages like English. Engine: (انجن), O-Level (پیراگراف) (O-لیول کے طلباء) Paragraph

2.5 Initial letter Capitalization

In English proper nouns, start of sentences can be recognized with the words with letter capitalized. For instance,

- The world is shrinking [initial letter of “The” is capitalized at the beginning of sentence]
- The delegation will meet Abu Bakar at Islamabad [initial letter of proper nouns “Abu Bakar” and “Islamabad” are capitalized.

[proper nouns starting with a normal character]. وفد ابوبکر سے اسلام آباد میں ملاقات کرے گا۔

2.6 Grammatical Words

Grammatical words are prepositions, adverbs, conjunctions etc. they themselves have not a very clear meaning. However, these words are used to complete sentences and there meanings are expressed in dictionaries with the help of examples. For instance, for, with, the, of, etc. In Urdu the examples of grammatical words are, سے ، نے ، کا، etc. say, nay, kaa respectively.

3 Literature Review

3.1 Historic perspective of Spell Checking

Wherever there is text processing, misspelled words come across, and these misspelled words are required to be detected and corrected. Thus research in spell detection and correction started in the period when text processing become common for computer users. In 1964, Fred J Damerau, articles (Fred J. Damerau 1964) explained fundamentals techniques for spelling detection and correction. According to Damerau, spell checking is a process of comparing an input index with a master list of acceptable terms, and rejects those word from the input which has no match in the master list. In tests conducted by Damerau, indicated that 80% of the spelling errors falls in single letter error, these errors are: *Substitution*; a letter is wrongly substituted by another letter,

Insertion; a letter is wrongly inserted at some position,

Deletion; a letter is wrongly deleted from some position,

Transposition; two letters are wrongly transposed.

Few examples words that are taken from the test data of Damerau are given in a Table 1.

| Error type | Correct word | Misspelled word |
|---------------|--------------|-----------------|
| Substitution | Absorbent | Absorbant |
| Insertion | Commitment | Committment |
| Deletion | Governmentt | Government |
| Transposition | Wierd | Weird |

Table 1: Extracts from Damerau Test data In a test conducted by Damrau, a data of 964 spell errors was taken for conducting a test. The results are presented in Table 2.

| | Substitution | Deletion | Transposition | Insertion | Multiple errors | Total |
|------------------------|--------------|----------|---------------|-----------|-----------------|-------|
| Correctly identified | 549 | 143 | 23 | 97 | 0 | 812 |
| Incorrectly identified | 18 | 10 | 0 | 2 | 0 | 30 |
| Not identified | 0 | 0 | 0 | 0 | 122 | 122 |
| | | | | | | |
| Total | 567 | 153 | 23 | 99 | 122 | 964 |

Table 2: Major error types for spelling errors

3.2 Spelling Error classification

Spelling errors are classified into two types, namely typographic errors and cognitive errors (Kyongho Min, William H. Wilson, Yoo-Jin Moon), (Tahira Naseem 2004). Typographic errors are those errors in which the person typing the text knows spelling, however mistype the word. For instance, a user intends to type “listen” but wrongly types “listyen”. The additional adjacent key ‘y’ is pressed while the typist intended to type ‘t’ in the word “listen”. Thus, the “listyen” example pertains to insertion as explained by Damerau. In case of cognitive error, spelling of word is not known to the writer. Due to existence of homophone alphabet set in Urdu language, cognitive errors are found in the Urdu written text. For instance [ض، ذ]

For Urdu spelling errors trends, a study was conducted by (Tahira Naseem et al). The data for the study was taken from newspaper text, and students term papers. In English, [s, c] are phonetically equivalent, or termed as homophone alphabets; given in example [race, rase]. In case of Urdu text, there are several homophones characters sets. For instance [ذ، ز، ض]. Similarly,

visually similar character also exists in Urdu text [ذ، ذ]. Their study exhibited the results illustrated in Table 3.

| Newspaper Text | | | |
|----------------|--------------|------------------|----------------------|
| | Total errors | Visually Similar | Phonetically similar |
| Substitution | 75 | 40 | 12 |
| Deletion | 42 | 4 | 5 |
| Insertion | 21 | 2 | 1 |
| Transposition | 12 | 3 | 0 |
| Total | 150 | 49 | 18 |

Table 3: Single Edit Distance Errors in Urdu

3.3 Spelling Error correction techniques

The spelling correction solution comprises of three phases:

- Detection of Spelling error
- Finding candidate word(s) for the misspelled words
- Order candidate according to match strength

The following are the techniques that are employed by various spell correction tools in many languages.

- Minimum Edit Distance technique
- Similarity Key technique
- Neural Networks technique

3.3.1 Minimum Edit Distance Technique

In 1965, the Minimum Edit Distance technique was given by Vladimir Levenshtein, to compute minimum edit distance or edit operation required to transform one string *str1* to another string *str2*. In this technique, a matrix is taken of dimension $m \times n$, where m , and n represents the length of two strings *str1*, and *str2*. One of the string say *str1*, is placed at the top row of matrix, and the *str2* is placed at leftmost column. On execution of Edit Distance algorithm, each cell of the matrix is filled with the difference of edit operations performed.

The Minimum Edit Distance algorithm measures distance between two strings. An *insertion* operation takes place, when a alphabet is inserted in a non-word sequence to make it correct word. Similarly, *deletion* operation takes place, when a alphabet is deleted from a non-word sequence to make it correct word, *substitution* operation takes place when a alphabet is substituted in a non-word sequence to make it correct word. If there are w number of words in a lexicon, the minimum edit distance algorithm performs w comparisons of a misspelled word with all w words in dictionary. To minimize the comparisons, re-

verse edit distance algorithm (Eric Brill and Robert C. Moore), (M. D. Kernighan, K. W. Church, and W. A. Gale 1990) is proposed. In modified algorithm only $53n + 25$ words comparisons are performed, where n is length of misspelled English word.

3.3.2 Similarity Key Technique

This technique is based upon the key assignment to the words of a language. Words are composed of alphabet. In this technique a set of alphabet is taken based on sound similarity. It is to be noted that the key is not unique, and will be explained shortly. The following are the sets in Similarity Key technique.

| Digit of Key | Alphabet |
|--------------|-----------------|
| 0 | a,e,i,o,u,h,w,y |
| 1 | b,f,p,v |
| 2 | c,g,j,k,q,s,x,z |
| 3 | d,t |
| 4 | L |
| 5 | m, n |
| 6 | R |

Table 4: Similarity keys for English Alphabets

In this technique, the key is calculated for a misspelled word. Then the words from the lexicon that have the same key value are extracted for candidature of a correctly spelled word. The key are generated by keeping the first letter of the word followed by digits mapped from the Table 4. For instance the key t0140 is generated for the word *table*. The zero and are eliminated from the key. Similarly repetition of a character is collapsed. Thus in second step the key t0140 becomes t14 for the word *table*.

The Urdu character set is also composed of many homophones. A study has been conducted (Tahira Naseem 2004) on spelling mistakes of Urdu words in context to soundex.

The similar Urdu sounded letter are shown in Table 5, and Table 6 of soundex scheme 0-F, and 0-9 respectively.

| code | Alphabet |
|-------|----------|
| 0 | س،ش،ص،ث |
| 1 | ت،ط،ٹ |
| 2 ~ D | ----- |
| E | ل |
| F | و |

Table 5: Similar sounded letters – Urdu (Scheme 0 ~ F)

| code | Alphabet |
|------|----------|
| 0 | س،ش،ص،ث |
| 1 | ت،ط،ٹ |

| | |
|-------|-------|
| 2 ~ 8 | ----- |
| 9 | ل |

Table 6: Similar sounded letters – Urdu (Scheme 0 ~ 9)

3.3.3 Neural Networks Technique

Neural networks are used in environments where systems are trained on specific error patterns (O. Matan, C. J. C. Burges, Y. LeCun, and J. S. Denker 1992). Thus neural networks can be used for spell correction. The neural network is trained in for spell errors for a specific domain in which the spell correction system will be used. The back propagation method is commonly used in neural network training (V. J. Hodge and J. Austin 2003). The method comprised of three layers; the input, hidden, and output layer. The nodes from inner to output layer are connected through a link through hidden layer. In the training phase, weights are computed and assigned to the nodes from input layer to output layer. The weights represent the relation between the nodes.

4 Proposed Technique for Urdu Spell Checking - Reverse Edit Distance

It is obvious that q linear comparisons are required for a misspelled word with all words of lexicon containing q words. 70,000 words lexicon, requires 70,000 comparisons. Reverse edit distance technique is proposed (M. D. Kernighan, K. W. Church, and W. A. Gale 1990) in which permutation of edit distance one are generated and compared with lexicon words which are in alphabetic order. Urdu has 42 characters, and for a Urdu word of length n , a total of $86n + 41$ strings are checked, which is far below than lexicon words q . The break-up is as under:

| | |
|--|------------|
| Insertions of ا، ب، .. ی، ے at $n+1$ positions in a word. | $42(n+1)$ |
| Deletion of one alphabet in turn from word. n | |
| Substitution of ا، ب، .. ی، ے in turn at each position in word | $42n$ |
| Transpositions of adjacent alphabets in word | $n-1$ |
| ----- | ----- |
| Total comparisons | $86n + 41$ |
| ===== | ===== |

The reverse edit distance technique can be employed for Urdu Spell system. The usage of this technique is selected due to its efficient approach.

Urdu misspelled word $w =$ 'نہکی'

Substitutes:

The first letter 'ن' of Urdu misspelled word نہک is substituted with ے، ی،، ب، ا in turn resulting:

ےہک،، تہک، پہک، بہک، اہک

Likewise, The second letter 'ہ' of misspelled word نہک is substituted with alphabet ے، ی،، ب، ا in turn resulting: ےنک،، نمک،، نیک، نک، نتک، نپک، نیک، ناک

Likewise, The third, and fourth letter get substituted.

Inserts :

In next set of permutations, characters ے، ی،

.....، ب، ا are inserted in turn at position 1,2,3,

and 4 of the misspelled word نہک.

ینہک،، تتہک، پتہک، بہتہک، انہک

Transposes

نکھ، ہنک

Deletes

نہنک، ہک

4.1 Dictionary lookup

Finite State Automata (FSA) are based on alphabets U of a language L. A string S, comprises of alphabet from U. If S belongs to U, then there will be a path from the initial state to the final state of the Finite state automata, else will declare misspelled word. If a correct word that is not available in the lexicon would be required to be added while the spell checking application highlight as a misspelled word.

4.2 Edit Distance (Two operations)

Despite the fact that misspelled words are corrected by one edit operation. Thus it may happen that out of $86n + 41$ permutations generated from an Urdu non-word may lead to zero match in the lexicon. Thus in our proposed work, edit distance is procedure is called again on all the permutation generated by the edit distance one operation. This drastically slows down the system as permutations (2^{nd} cycle) for each of the permutation generated at edit distance 1 will be generated.

4.3 Candidate Generation

Candidates are selected upon the existence of any match of the permuted word in lexicon. In our example case two candidates are generated: $c = [\text{'نمک'}, \text{'نیک'}]$

4.4 Reverse Edit Distance Efficiency

The Reverse Edit Distance algorithm is grossly better than the conventional Edit Distance algo-

rithm. This can be explained with simple calculation on data.

| | |
|------------------------------|------------------------------|
| Lexicon size (q) = | 100,000 words |
| Misspelled word length (n) = | 6 letters |
| | Edit Distance |
| | Reverse Edit Distance |
| Comparisons | 100,000 |
| | $86n + 41$ |
| | $= 86(6) + 41$ |
| | $= 557$ |

Thus in this example the reverse edit distance performs 180 times better as compare to edit distance technique.

4.5 Methodology Digest

This work presents spell checking for Urdu non-words employing lexicon lookup. Spell checkers are not available in the Urdu text processors, and the most important among them is the widely used Urdu Word processor still lacking the spell checking and correcting feature.

The Levenshtein distance or most commonly minimum edit distance algorithm is used as a basic and acceptable technique for spell correction. In our work, the same algorithm has been selected for getting candidate words for an Urdu non-word.

The working model of the proposed system is kept simple, being the system is build on basic works, as very scarce resources are available on Urdu Language and specifically the Urdu spell checking and correction. A three step approach has been employed in the system: Lexicon lookup, candidates' generation, and ranking candidates.

The system uses an lexicon or corpus that contain the correctly spell words, and this corpus is referred by system for ensuring that the given word is misspelled word or otherwise If match of input word is found in underlying corpus then word is correctly spelled, and declared misspelled if not found. For misspelled word, the process of finding correctly spelled word candidate starts. At this step, the Levenshtein distance or most commonly minimum edit distance algorithm is used to get the candidates. Most of the errors are at distance one. Damerau has observed that 80% of the error lies at edit distance one, that the 80% of the misspelled words need one edit operation of substitution, insertion, deletion, and transposition. For generation of candidates, edit distance one, and edit distance two has been used. The union of results of both edit distance one and edit distance two is taken to produce a list of candidate words. In the third and last step, best probable correctly spelled word for the mis-

spelled word based on the frequency of candidates word exists in the underlying corpus. The methodology is explained in the illustration:

| Misspelled word | Corpus | Candidate words | Top ranked word |
|-----------------|-------------|---|-------------------|
| ادر | Urdu corpus | ادا (34) اثر (23) اثر (2) صدر (40) اگر (12) | صدر get selected. |

Table 7. Illustration explaining methodology

In the above illustration, for the misspelled Urdu word ادر we have taken few candidate words ادا، اثر، اثر، صدر، اگر shown in the third column of Table 7 which are at edit distance one, that is one edit operation we can get these words from the misspelled word. Now to decide which of these candidate words is intended correct word that typist wanted to write cannot be flatly decided.

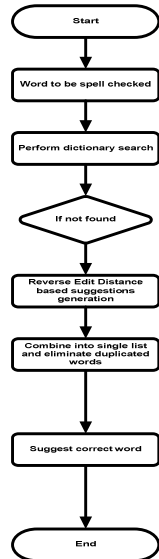


Figure 1. Flow chart depicting the process of spell checking and correction

5 Results and Discussions

5.1 Training Corpus

Two corpuses of 54,440 words, and 56,142 words [Center For Research In Urdu Language Processing] are taken that has been used as training lexicon. In subsequent work later, a comprehensive corpus will be taken for results. The training data is used for frequency count of the words. The frequency count is used for analysis of candidate words from a misspelled word. Let's take a simple corpus of 20 Urdu words to understand the training set concept:

ابتدائی نتائج ایک سو سال سے کچھ پہلے بتائے گئے تھے مگر تازہ ترین نتائج پچھلے سال اور اس سال سے پہلے

In the work presented, small hypothetical data is taken to understand the concept. In future, with implementation of algorithms, accuracy will be also bw calculated.

The frequency of these words is shown in the Table 8

| الفاظ (Words) | تعداد (Frequency) |
|---------------|-------------------|
| ابتدائی | 1 |
| نتائج | 2 |
| ایک | 1 |
| سال | 3 |
| سو | 1 |
| سے | 2 |
| کچھ | 2 |
| پہلے | 1 |
| بتائے | 1 |
| گئے | 1 |
| تھے | 1 |
| مگر | 1 |
| تازہ ترین | 1 |
| پچھلے | 1 |
| اور | 1 |
| اس | 1 |

Table 8. Frequency – Mini training set data

Now taking misspelled word سق, let the spell correction system generate two candidates; the word سو [frequency count = 1], and the word سے [frequency count = 2]. Based on frequency, suggested word is calculated which comes to be سے for the misspelled word سق

Taking another example of few misspelled words and executing our algorithm on a training set corpus of 54,400 words. We would see abstract behavior of algorithm and make a little discussion on the result generated. The following non-words are provided to the system for generation of candidate words, there after highlighting one word.

5.2 Urdu Non-words example

استعمال کٹی جانپی تی میں جس جس کششی جسگہ ایشک
دوسٹری قشسم کشے فلشٹر

Test word –i استعمال

Candidate word(s) استعمال

Suggested word استعمال

=====
Test word- ii کٹی

Candidate word(s) کسی کئی کری
Suggested word کسی

Test word - iii جاتی

Candidate word(s) جاتی ساتھی
Suggested word جاتی

Test word -iv میں

Candidate word(s) میں ملیں نہیں
Suggested word میں

Test word-v جس

Candidate word(s) جس باس جدا
Suggested word جس

Test word - vi کششی

Candidate word(s) کرتی کسری
Suggested word کرتی

Test word-vii جگہ

Candidate word(s) جگہ جبکہ
Suggested word جبکہ

Test word-viii کٹے

Candidate word(s) کرے کٹے
Suggested word کرے

Test word-ix فلٹر

Candidate word(s) فلٹر فوسٹر
Suggested word فلٹر

In the above example, we have taken 09 non-words of Urdu and pass through the reverse edit distance algorithm. Candidate words in the range of one to three words are generated for each of the non word. These are words are at edit distance of one or two from the corresponding non-word. The candidate words are generated by in-

sertions, deletion, substitution, or transposition operations.

5.3 Correct word suggestion

In the corpus the $c(\text{جاتی}) = 22$, and $c(\text{ساتھی}) = 5$, Thus $c(\text{جاتی}) > c(\text{ساتھی})$ thus the word جاتی is suggested as the top ranked word for the misspelled word $w = \text{جاتی}$

Insertion

The following are selected non-words examples specifically for **insertion** of alphabet that are passed through the proposed system to get candidate results:

Test word اصلاحت

Candidate word(s) اصلاحات [Other candidates are removed]

Action: ا is inserted between ح and ت

Deletion

The following are selected non-words examples specifically for **deletion** of alphabet that are passed through the proposed system to get candidate results:

Test word الجتھاؤ

Candidate word(s) الجھاؤ

Suggested word الجھاؤ

Substitution

The following are selected non-words examples specifically for **Substitution** of alphabet that are passed through the proposed system to get candidate results:

Test word قیسا

Candidate word(s) ایسا میرا پیدا
قیاس نیسب جیسا لینا

Suggested word جیسا

Transposition

The following are selected non-words examples specifically for **Transposition** of two adjacent

alphabets that are passed through the proposed system to get candidate results:

| | | | |
|-------------------|------|------|-----|
| Test word | بند | | |
| Candidate word(s) | بندر | بلند | بند |
| Suggested word | بلند | | |

We have noticed that the transposition errors are poorly corrected using reverse edit distance as compared to edit distance algorithm. In above results, the words قبیل, پچھیے, تبدل are not properly corrected by the algorithm.

6 Conclusion

Urdu language has rich literature, spoken in south asia, however having scarce resources in context of computer based applications. In this work, focus is on spelling error detection and correction feature in these electronic applications. Our this work is concentrated on gathering various spell checking and correction techniques that are suitable for correcting Urdu spelling errors . Reverse Edit Distance algorithm complexity is computed to be $86n + 41$. The algorithm has been implemented in other languages like English, and has to be implemented for Urdu.

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