Translation Memory Systems Have a Long Way to Go

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Abstract

The TM memory systems changed the work of translators and now the translators not benefiting from these tools are a tiny minority. These tools operate on fuzzy (surface) matching mostly and cannot benefit from already translated texts which are synonymous to (or paraphrased versions of) the text to be translated. The match score is mostly based on character-string similarity, calculated through Levenshtein distance. The TM tools have difficulties with detecting similarities even in sentences which represent a minor revision of sentences already available in the translation memory. This shortcoming of the current TM systems was the subject of the present study and was empirically proven in the experiments we conducted. To this end, we compiled a small translation memory (English-Spanish) and applied several lexical and syntactic transformation rules to the source sentences with both English and Spanish being the source language.

The results of this study show that current TM systems have a long way to go and highlight the need for TM systems equipped with NLP capabilities which will offer the translator the advantage of he/she not having to translate a sentence again if an almost identical sentence has already been already translated.

1. Introduction

While automatic translation has taken off to work reasonably in some scenarios and to do well for gisting purposes, even today, against the background of the latest promising results delivered by statistical Machine Translation (MT) systems such as Google Translate and latest developments in Neural Machine Translation and in general Deep Learning for MT, automatic translation gets it often wrong and is not good enough for professional translation. Consequently, there has been a pressing need for a new generation of tools for professional translators to assist them reliably and speed up the translation process. Historically, it was Krollman who first put forward the reuse of existing human translations in 1971. A few years later, in 1979 Arthern went further and proposed the retrieval and reuse not only of identical text fragments (exact matches) but also of similar source sentences and their translations (fuzzy matches). It took another decade before the ideas sketched by Krollman and Arthern were commercialised as a result of the development of various computer-aided translation (CAT) tools such as Translation Memory (TM) systems in the early 1990s. These translation tools revolutionised the work of translators and the last two decades saw dramatic changes in the translation workflow.

The TM memory systems indeed changed the work of translators and now the translators not benefiting from these tools are a tiny minority. However, while these tools have proven to be very efficient for repetitive and voluminous texts, they operate on fuzzy (surface) matching mostly and cannot benefit from already translated texts which are synonymous to (or paraphrased versions of) the text to be translated. The match score is mostly based on character-string similarity, calculated through Levenshtein distance (“measure of the minimum number of insertions, deletions and substitutions needed to change one sequence of letters into another.”; Somers 2003)

The limitations of the traditional TM systems in terms of matching have been highlighted by a number of authors. By way of example, Somers (2003) gives the example below to illustrate one drawback:

a. Select ‘Symbol’ in the Insert menu.
b. Select ‘Symbol’ in the Insert menu to enter character from the symbol set.
c. Select ‘Paste’ in the Edit menu.
Given the input sentence (a), (c) would be the highest match percentage, as it differs in only two words, while (b) has eight additional words. Intuitively (b) is a better match since it includes the text of (a) in its entirety.

Before Mitkov (2005) elaborated on the specific matching limitations of TM systems, Macklovitch and Russel (2000) pointed out ‘Current Translation Memory technology is limited by the rudimentary techniques employed for approximate matching’. They went on to illustrate that unless an TM system can do morphological analysis, it will have difficult recognising that (f) is more similar to (d) than (e):

d. The wild child is destroying his new toy 

f. The wild children are destroying their new toy

Gow (2003) notes that SDL Trados gives the segments ‘Prendre des mesures de dotation et de classification,’ and ‘Connaissance des techniques de rédaction et de révision.’ a match rating of as high as 56% even though the above sentences have nothing to do with each other the reason being that because half of the words are the same and they are in the same position, even though the common words are only function words.

Recent work on new generation TM systems (Gupta 2015; Gupta et al. 2016a; Gupta et al. 2016b; Timonera and Mitkov 2015; Gupta and Orasan 2014) show that when NLP techniques such as paraphrasing or clause splitting are applied, TM systems performance is enhanced.

While that it is clear that TM systems are incapable of any linguistic or semantic interpretation, we maintain that they have difficulties with detecting similarities even in sentences which represent a minor revision of sentences already available in the translation memory. In order to substantiate this claim empirically, we conducted the following experiments.

We conducted experiments using a small translation memory (English-Spanish) in which we apply several lexical and syntactic transformation rules to the source sentences – both for English and Spanish serving as source language. The transformation rules are selected in such a way that they are simple and the transformed sentences do not change in meaning. The hypothesis of this study is that in many cases the TM systems cannot detect the fact that the transformed sentences are practically the same as sentences already translated as the match computed is below the threshold. This in turn, would mean insufficient efficiency as the new, practically the same sentences, will have to be translated again. For the purpose of this study we experimented with the well-known TM systems Trados, Wordfast, Omega T and MemoQ.

2. Data Description

For the purpose of this experiment a translation memory or alternatively parallel corpora were needed. To this end, we compiled a parallel English-Spanish corpus consisting of online documents of the European Union and the United Nations. The English documents selected were Charter of the Fundamental Rights, Declaration on the Right to Development, Declaration on the Rights of Persons Belonging to National or Ethnic, and Religious and Linguistic Minorities, United Nations Declaration on the Rights of Indigenous People, and Universal Declaration of Human Rights. We also selected the Spanish equivalents of these documents: Carta de los derechos fundamentales de la Unión Europea, Declaración de las Naciones Unidades sobre el derecho al desarrollo, Declaración sobre los derechos de las personas pertenecientes a minorías nacionales o étnicas, religiosas y lingüísticas, Declaración de las Naciones Unidades sobre los derechos de los pueblos indígena and Declaración universal de los derechos humanos.1 The size of the English corpus was 14,153 words while the size of the Spanish corpus is 15,461 words (for more details see Table 1).

1 The documents in English and Spanish are identical in contents even though the titles are slightly different.
We are aware the compiled corpus is very small but we regard this study and results as preliminary. In fact for the purpose of our experiments we selected a small sample of 150 aligned sentences in English and Spanish; this ‘experimental TM’ served as a basis for the experiments outlined.

### 3. Transformations

For the purpose of this study, we developed 10 transformation rules. Rule 1 was to transform an original sentence in active voice into a passive voice sentence, if possible. Rule 2 was a mirror image of rule 1 – transform a sentence in passive voice into active voice sentence. Rule 3 had to do with changing the order inside the sentence – changing the order of words, phrases or clauses within a sentence. Rule 4 sought to replace a word with a synonym whereas Rule 5 replaced 2 words of a sentence with their synonyms. The replacement with synonyms was applied to nouns, verbs and adverbs and in cases where the existence of synonym with identical meaning was ‘obvious’. Rule 6 built on rule 3 by changing the order within a sentence but in addition replaced a noun with a pronoun for which it served as antecedent. Rule 7 was a combination of rule 1 and 4 – change of active voice into passive and replacement of a word with its synonym. Rule 8 changed passive voice into passive and like rule 6 - noun with a coreferential pronoun. Rule 9 was combination of rule 3 and rule 5, and finally was a subsequent application of rule 2, rule 3 and pronominalisation of a noun.

Table 2 below lists the rules with examples in English, whereas Table 3 lists the rules with examples in Spanish.

| Rule | Transformation | Original sentence (English) | Transformed sentence (English) |
|------|----------------|-----------------------------|-------------------------------|
| 1    | Change active to passive voice | States must take measures to protect and promote the rights of minorities and their identity. | Measures must be taken to protect and promote the rights of minorities and their identity. |
| 2    | Change passive to active voice | The history, traditions and cultures of minorities must be reflected in education. | The education must reflect the history, traditions and cultures of minorities. |
| 3    | Change word order, phrase order or clause order | Reaffirming those indigenous peoples, in the exercise of their rights, should be free from discrimination of any kind. | Reaffirming those indigenous peoples should be free from discrimination of any kind, in the exercise of their rights. |
| 4    | Replace one word with its synonym | Everyone has the right to | Every person has the right to |
Replace two words with its synonym

| Rule | Transformation | Original sentence (Spanish) | Transformed sentence (Spanish) |
|------|----------------|-----------------------------|--------------------------------|
| 5    | Change active to passive voice AND change word order, phrase order or clause order | Las sociedades de todo el mundo disfrutan de la diversidad étnica, lingüística y religiosa. | Las sociedades mundiales disfrutan de la variedad étnica, lingüística y religiosa. |
| 6    | Change active to passive voice AND replace one word into a pronoun | Todos los niños, nacidos de matrimonio o fuera de matrimonio, tienen derecho a igual protección social. | Ellos tienen derechos a igual protección social, nacidos de matrimonio o fuera de matrimonio. |
| 7    | Change active to passive voice AND replace one word with its synonym | Los Estados tomarán las medidas que sean necesarias para lograr progresivamente que este derecho se haga plenamente efectivo. | Las decisiones que sean necesarias para lograr progresivamente que este derecho se haga plenamente efectivo serán tomadas por los Estados. |
Table 3: Transformation rules and examples in Spanish

4. Experiments, Results and Discussion

We use the above parallel corpus as a translation memory and experiment with both English and Spanish as source languages. If we had to translate again a sentence from the source Language, the match would be obviously 100%. For the purpose of the experiment, each sentence of the source text undergoes a transformation after applying the rules listed below, which convert the original sentences into syntactically different but semantically same sentences. As the new sentences have the same meaning, it would be desirable that the TM systems produce a high match between the transformed sentences and the original ones. By ‘high match’ we mean a match above the threshold of a specific TM tool so that the user can benefit from the translation of the original sentence being displayed.

| Rule | # Sentences | Trados <75% | Trados Failure % | Wordfast <75% | Wordfast Failure% | OmegaT <75% | OmegaT Failure% | MemoQ <75% | MemoQ Failure% |
|------|-------------|-------------|-----------------|---------------|-----------------|-------------|----------------|-------------|----------------|
| 1    | 28          | 10          | 35.71           | 2             | 7.14            | 14          | 50             | 11          | 39.28          |
| 2    | 27          | 7           | 25.92           | 6             | 22.2            | 14          | 51.85          | 13          | 48.15          |
| 3    | 84          | 8           | 9.52            | 0             | 0               | 18          | 21.43          | 24          | 28.57          |
| 4    | 150         | 2           | 1.33            | 21            | 14              | 6           | 4              | 13          | 8.6            |
| 5    | 150         | 10          | 6.67            | 77            | 51              | 7           | 4.6            | 62          | 41.3           |
| 6    | 29          | 11          | 37.93           | 5             | 17.24           | 14          | 48.27          | 16          | 55.17          |
| 7    | 26          | 9           | 34.61           | 14            | 53.85           | 11          | 42.31          | 22          | 84.61          |
| 8    | 9           | 2           | 22.22           | 3             | 33.3            | 5           | 55.55          | 6           | 66.67          |
| 9    | 84          | 22          | 26.19           | 42            | 50              | 42          | 50             | 64          | 76.2           |
| 10   | 20          | 4           | 20              | 5             | 25              | 9           | 45             | 14          | 70             |

Table 4: Matching results English

The results obtained with English as a source language show that the lexical and syntactic transformations cause significant problems to the TM systems and their inability to return matching above the default threshold means that the translator may miss out on the re-use of translations he/she has already made. In the case of Trados there are as many as 36% of the sentences not being returned after being transformed with rule 1; this figure goes up to 38% when applying rule 6 and 35% with rule 7. Wordfast fails to propose similarity for 54% of the sentences once rule 7 is applied. As for Omega T, the application of rule 8 results in
56% of the transformed sentences not being detected as similar to the original sentence. MemoQ’s failures are even more dramatic in that the applications of rules 10, 9 and 7 leads to inability to detected similarity in 70%, 76% and 85% of the cases respectively! Overall, MemoQ reports the most negative results which is surprising given the very positive feedback of users in general for this tool. It appears that TM systems in general have more problems with syntactic transformations – after applying rules 1, 2, 3 or the combined rules 6-10. The TM systems also report fewer problems for lexical transformation only – e.g. after applying rules 4 and 5 only.

As the transformations are language specific, we conducted similar experiments with Spanish being the source languages. We transformed the original Spanish sentences using the above rules and the TM systems computed the match between the transformed sentences and the original ones.

| Rule | # Sentences | Trados | Wordfast | OmegaT | MemoQ |
|------|-------------|--------|----------|--------|-------|
|      |             | <75% Failure% | <75% Failure% | <75% Failure% | #<75% Failure% |
| 1    | 50          | 10      | 20       | 7      | 14    | 22  | 44  | 12  | 24  |
| 2    | 18          | 11      | 61.1     | 10     | 55.55 | 11  | 61.11 | 12  | 66.67 |
| 3    | 91          | 20      | 21.98    | 1      | 1.10  | 19  | 20.88 | 23  | 25.27 |
| 4    | 150         | 5       | 3.33     | 24     | 16    | 5   | 5    | 14  | 9.33  |
| 5    | 150         | 13      | 8.67     | 95     | 63.3  | 33  | 22   | 55  | 36.67 |
| 6    | 44          | 18      | 40.9     | 5      | 11.36 | 16  | 36.36 | 20  | 45.45 |
| 7    | 50          | 16      | 32       | 29     | 58    | 19  | 38   | 24  | 48    |
| 8    | 17          | 10      | 58.82    | 8      | 47.05 | 10  | 58.82 | 9   | 52.94 |
| 9    | 91          | 36      | 39.56    | 50     | 54.94 | 41  | 45.05 | 55  | 60.44 |
| 10   | 25          | 6       | 24       | 9      | 36    | 9   | 36   | 18  | 72    |

Table 5: Matching results Spanish

The results obtained with Spanish as a source language show that the lexical and syntactic transformations cause more significant problems to the TM systems than with English as a source language. In the case of Trados there are as many as 40% of the sentences not being returned after being transformed with rule 6 and 9; this figure goes up around 60% when applying rule 8 and 61% with rule 2. Wordfast fails to propose similarity for around 55% of the sentences once rule 2 and 9 are applied; and up to 63% when rule 5 is also applied. As for Omega T, the application of rule 8 results in 59% of the transformed sentences not being detected as similar to the original sentence and up to 61% when applying rule 2. MemoQ’s failures after applying rules 8, 9 and 2 leads to inability of detecting similarity in 52%, 60% and 67% of the different cases. On the whole, every TM system exhibits retrieval failures mostly related to combining different rules.

It is worth noting the high errors rates for all TM systems even with rule 2 which confirms their inability to deal with syntactic transformations.

Finally, it is worth observing that for Spanish as source language the matching failures are higher. As the above TM systems have no NLP functionalities and usually use Levenstein distance as a matching algorithm, we conjecture that this has to do with the slightly more complex syntax in Spanish.

5. Conclusion

Current TM systems have a long way to go. The above results highlight the need for TM
technology to embrace NLP techniques such as parsing or paraphrasing. A TM system equipped with NLP capabilities will offer the translator the advantage of he/she not having to translate a sentence again if an almost identical sentence has already been already translated.

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