Building an Arabic Machine Translation Post-Edited Corpus: Guidelines and Annotation

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Abstract
We present our guidelines and annotation procedure to create a human corrected machine translated post-edited corpus for the Modern Standard Arabic. Our overarching goal is to use the annotated corpus to develop automatic machine translation post-editing systems for Arabic that can be used to help accelerate the human revision process of translated texts. The creation of any manually annotated corpus usually presents many challenges. In order to address these challenges, we created comprehensive and simplified annotation guidelines which were used by a team of five annotators and one lead annotator. In order to ensure a high annotation agreement between the annotators, multiple training sessions were held and regular inter-annotator agreement measures were performed to check the annotation quality. The created corpus of manual post-edited translations of English to Arabic articles is the largest to date for this language pair.

Keywords: Post-Editing, Guidelines, Annotation

1. Introduction

In recent years, machine translation (MT) became widely used by translation companies to reduce their costs and improve their speed. Therefore, the demand for quick and accurate machine translations is growing. Machine translation (MT) systems often produce incorrect output with many grammatical and lexical choice errors. Correcting machine-produced translation errors, or MT Post-Editing (PE) can be done automatically or manually. Successful automatic post-editing approaches using manually corrected MT output were used by Elming (2006) and Simard et al. (2007). The availability of annotated resources is required for such approaches. When it comes to the Arabic language, to the best of our knowledge, there is no manually post-edited MT corpora available to build such systems. Therefore, there is a clear need to build such valuable resources for the Arabic language.

In this paper, we present our guidelines and annotation procedure to create a human corrected MT corpus for the Modern Standard Arabic (MSA). The creation of any manually annotated corpus usually presents many challenges. In order to address these challenges, we created comprehensive and simplified annotation guidelines which were used by a team of five annotators and one lead annotator. In order to ensure a high annotation agreement between the annotators, multiple training sessions were held and regular inter-annotator agreement (IAA) measures were performed to check the annotation quality. To the best of our knowledge, this is the first published machine translation manual post-editing annotation effort for Arabic of this scale.

In the next sections, we review related work (Section 2), describe our corpus and the development of the guidelines (Sections 3-4), and present our annotation procedure (Section 5), than we present the annotation evaluation in Section 6, finally we conclude our work in Section 7.

2. Related Work

Large scale manually corrected MT corpora are not yet widely available due to the high cost related to building such resources. Wisniewski et al. (2014) created a corpus of machine translation errors extracted from several translation students taking part in a master program in specialized translations. The texts are translated from English to French. A portion of the corpus contains an analysis of the type of errors made by the MT system. Elming (2006) created a 265K-word English-Danish MT manually corrected corpus by a human professional translator. The full corpus covers the chemical patents domain. Simard et al. (2007) created a 500K-word corpus of manually edited French-English and English-French MT from the Canadian Job Bank website. The corpus is a collection of blocks composed of the source language texts, the machine translation output of a rule-based MT system and the final post-edited version done by a human translator. Moreover, Avramidis et al. (2014) built a corpus of human-annotated machine translations which was evaluated by professional human translators for the following three language pairs: German-English, English-German and Spanish-German.

Fishel et al. (2012) created a corpus of automatically produced translations with detailed manual translation error analysis of 576 sentences for four language pairs: English-Czech; French-German; German-English; English-Serbian. Popescu-belis et al. (2002) produced a small corpus of 50 texts translated by students and corrected by their professors and all translation errors are annotated with their corrections in this corpus. For Arabic, we cite the effort of Bouamor et al. (2014) who created a medium scale human judgment corpus of Arabic machine translation using the output of six MT systems and a total of 1892 sentences and 22K rankings.

Our corpus is a part of the Qatar Arabic Language Bank (QALB) project, a large scale manually annotated annotation project (Zaghouani et al., 2014b; Zaghouani et al.,...
We refer to Zaghouani et al. (2014b) for more details about these errors.

In the MT post-editing guidelines, we provide the annotators with detailed annotation procedure and explain how to deal with borderline cases in order to increase the inter-annotator agreement. We include many annotated examples to illustrate some specific cases of machine translation correction rules as seen in the example of Figure 3 showing the original English text, Figure 4 showing a sample machine translation output and Figure 5 showing a sample human post-edited version of the sample.

Since there are equally-accurate alternative ways to edit the machine translation output, all being considered correct, some using fewer edits than others, therefore, the task can be very challenging.

We explained in the guidelines that the machine translated texts should be corrected with a minimum number of edits necessary to achieve an acceptable translation quality. However, correcting the accuracy errors and producing a semantically coherent text is more important than minimizing the number of edits and therefore, the annotators were asked to pay attention to the following three aspects: accuracy, fluency and style.

**Accuracy**
The accuracy of the translation is very important and any missing translation should be added to ensure the coverage of the semantics of the source sentence. In all cases, the edited machine translation text should have the same meaning as the English source text. Word and phrases rearrangement over long distances is only permitted if it is really needed as the annotators are encouraged to use as much of the raw MT output as possible. Finally, the edited machine translation should not add any information that is not present in the English source text. In the same way, it should not omit any information or meaning that is in the English source text as seen in the following example in Table 1.\(^4\)

**Fluency**
Grammatical and agreement errors occur frequently in MT texts and they should be always corrected. Word reordering is only permitted when it is needed to correct the syntax or meaning in the sentence as seen in the example of Table 2.

**Style**
Since machine translation output is fully automatically generated, it has its own style that may seem unfamiliar or unnatural although, in some cases it has an acceptable word order and conveys the meaning of the English source text accurately. Therefore, in such cases the style of writing should not be modified or improved as can be seen in Table 3.

The MT post-editing annotation guidelines will be published as a technical report and will be available in the near future on the QALB project web page.\(^5\)

### 5. Annotation Procedure

The annotation team consisted of a lead annotator and five annotators. The lead annotator is also the annotation workflow manager of this project. He evaluates the quality of the

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1. Spelling errors: mostly letter Yaa and hamza errors in the MT texts.
2. **Word choice errors**: a very frequent error in texts produced by MT systems.
3. **Morphology errors**: mostly related to an incorrect inflection or derivation.
4. **Syntactic errors**: the MT systems used in this project produced many cases of wrong gender and number agreement and also errors related to definiteness and wrong case and tense assignment.
5. **Proper names errors**: we observed many cases of named entities being improperly translated into Arabic.
6. **Dialectal usage errors**: the dialectal is generally not present in the MT texts.
7. **Punctuation errors**: in some cases punctuation signs appear in the wrong place.

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\(^1\)El Kholy and Habash (2012) present some of the major challenges for statistical machine translation from English into Arabic.

\(^2\)https://en.wikinews.org

\(^3\)https://cloud.google.com/translate

\(^4\)Arabic transliteration is presented in the Habash-Soudi-Buckwalter scheme (Habash et al., 2007).

\(^5\)http://nlp.qatar.cmu.edu/qalb/
Japan has an extensive web of highways with thousands of tunnels.

Brazil’s Syrians divided over unrest.

### Table 1: Example of accuracy errors. Words in bold not present in the original text

| Original | Japan has an extensive web of highways with thousands of tunnels. |
|----------|-----------------------------------------------------------------|
| Machine Translation | أَلَا يَا بَنِي عَبَّاد ِعَلَى مَكْتوبِ الرَّجُلِ بَعْضَ مِنَ الْبَحْرِ. |
| Post-Edit | أَلَا يَا بَنِي عَبَّاد ِعَلَى مَكْتوبِ الرَّجُلِ بَعْضَ مِنَ الْبَحْرِ. |

| Original | swryw AlbrAzyl maqswn Hwl AIADTrAbAt. |
|----------|---------------------------------------|
| Machine Translation | السُوْرَى وَالْبَزْيُ الْأَنْطَراَبِ. |
| Post-Edit | السُوْرَى وَالْبَزْيُ الْأَنْطَراَبِ. |

### Table 2: Example of fluency error. Words in bold have the wrong word order.

| Original | Brazil’s Syrians divided over unrest. |
|----------|--------------------------------------|
| Machine Translation | السُوْرَى وَالْبَزْيُ الْأَنْطَراَبِ. |
| Post-Edit | السُوْرَى وَالْبَزْيُ الْأَنْطَراَبِ. |

All post-editing action history previously mentioned are recorded in a database and can be exported to an XML file. Figure 2 shows an example of how the annotation actions are stored in the XML annotation export file.

Finally, and in order to increase the post-editing speed and prior to the first human pass, an automatic post-editing pass is done through MADAMIRA (Pasha et al., 2014), a tool that automatically corrects common spelling errors using a prediction model based on the words in-context. MADAMIRA uses a morphological analyzer to produce, for each input word, a list of analyses specifying every possible morphological interpretation of that word, covering all morphological features of the word. Most of the errors automatically corrected are related to Ya/Alif-Maṣūra, Ha/Ta-Marbuta and Hamzated Alif forms, which are common spelling errors in Arabic.\(^6\)

### 6. Evaluation

#### 6.1. Inter-Annotator Agreement

We use Word error Rate (WER) as a proxy of the Inter-annotator agreement. If the WER of two different annotations of the same sentences is low, we assume there is a high agreement between them. To evaluate the MT post-editing quality, we measure the inter-annotator agreement (IAA) on randomly selected files to ensure that the annotators are consistently following the annotation guidelines. A high annotation agreement is a good indicator of the data quality.

6 For more information on Arabic orthography and other issues of Arabic NLP, see (Habash, 2010).
Figure 1: The MT Post-Editing annotation interface. Edited words are highlighted in blue. Inserted words are highlighted in purple.

```xml
<ACTION_HISTORY>
  <ACTION actionType="edit" annotatorID="23" newText="الأمريكي" passNum="1" tokenID="34" />
  <ACTION actionType="edit" annotatorID="23" newText="ال" passNum="1" tokenID="37" />
  <ACTION actionType="edit" annotatorID="23" newText="الآسي" passNum="1" tokenID="46" />
  <ACTION actionType="edit" annotatorID="23" newText="والأص" passNum="1" tokenID="106" />
  <ACTION actionType="edit" annotatorID="23" newText="ال" passNum="1" tokenID="107" />
  <ACTION actionType="edit" annotatorID="23" newText="الأمر" passNum="1" tokenID="112" />
  <ACTION actionType="edit" annotatorID="23" newText="ال" passNum="1" tokenID="132" />
  <ACTION actionType="edit" annotatorID="49" newText="ثصص" passNum="2" tokenID="6" />
  <ACTION actionType="edit" annotatorID="49" newText="وقا" passNum="2" tokenID="10" />
  <ACTION actionType="edit" annotatorID="49" newText="ال" passNum="2" tokenID="11" />
  <ACTION actionType="edit" annotatorID="49" newText="الوليات" passNum="2" tokenID="12" />
  <ACTION actionType="delete" annotatorID="49" passNum="2" tokenID="16" />
  <ACTION actionType="add_token_before" annotatorID="49" tokenID="21" />  
  <ACTION actionType="delete" annotatorID="49" passNum="2" tokenID="25" />
  <ACTION actionType="edit" annotatorID="49" newText="الي" passNum="2" tokenID="24" />
  <ACTION actionType="add_token_before" annotatorID="49" tokenID="30" />
  <ACTION actionType="move_before" annotatorID="49" targetTokenID="48" tokenID="37" />
  <ACTION actionType="edit" annotatorID="49" newText="الأمر" passNum="2" tokenID="37" />
  <ACTION actionType="edit" annotatorID="49" newText="ليس" tokenID="45" />
  <ACTION actionType="move_after" annotatorID="49" targetTokenID="49" tokenID="51" />
  <ACTION actionType="add_token_before" annotatorID="49" tokenID="137" />
  <ACTION actionType="edit" annotatorID="49" newText="ال" passNum="2" tokenID="58" />
  <ACTION actionType="edit" annotatorID="49" newText="نعم" passNum="2" tokenID="60" />
  <ACTION actionType="delete" annotatorID="49" passNum="2" tokenID="59" />
</ACTION_HISTORY>
```

Figure 2: Extract of output file showing the correction action history.
It's been five years since pro-democracy protests started.

انها كانت خمس سنوات منذ أن بدأت الاحتجاجات المؤيدة للديمقراطية.

قد مر خمس سنوات منذ أن بدأت الاحتجاجات المؤيدة للديمقراطية.

| Original | It's been five years since pro-democracy protests started. |
| Machine Translation | انها كانت خمس سنوات منذ أن بدأت الاحتجاجات المؤيدة للديمقراطية. |
| Post-Edit | لقد مر خمس سنوات منذ أن بدأت الاحتجاجات المؤيدة للديمقراطية. |

Table 3: Example of machine translation unnatural but acceptable style shown in the words in bold. No correction is needed in this case.

| Raw vs Gold | IAA(Round1) | IAA(Round2) |
|-------------|-------------|-------------|
| QALB L1 Corpus | 24.45 | 3.80 | N/A |
| QALB L2 Corpus | 37.64 | 14.67 | 3.35 |
| QALB MT Corpus | 31.75 | 16.87 | 4.92 |

Table 4: Comparison between the MT corpus and the L1 and L2 corpus with the percentage of changes from the RAW output against the gold output and the inter-annotator agreement (IAA) on all ‘words’ in terms of average WER (Punctuation is ignored). Round1 is basic IAA comparing two annotations starting from raw output text. Round2 starts with the output of Round1.

| Original | ‘Traditional museums are run by the old people.’ |
| Machine Translation | يتم تشغيل المتاحف التقليدية التي كتبها كبار السن. |
| Annotator A (Good) |المتاحف التقليدية تتشكل من طرف كبار السن. |
| Annotator B (Acceptable) | كبار السن يديرون المتاحف التقليدية. |
| Annotator C (Bad) | تدار المتاحف التقليدية من قبل كبار السن. |

Table 5: Example of multiple post-editing corrections of an MT sentence

The IAA is measured over all pairs of annotations to compute the AWER (Average Word Error Rate). In this evaluation, the WER measures the post-editing errors against all words in the text, the lower the WER between two annotations, the higher is their agreement (Snover et al., 2006).

The IAA results shown in Table 4 include the results obtained in the current work as well as the results from the previous work described in Zaghouani et al. (2014b) for L1 corpus and in Zaghouani et al. (2015) for L2 corpus. We included the results from previous work to be able to compare IAA scores across the different genres.

The IAA results for the MT corpus are computed over 20 files (2,980 words) post-edited by at least three different annotators for the MT corpus and over 200 files (10,288 words) for the L1 corpus and finally 20 files (3,188 words) for the L2 corpus.

Table 4 shows the number of changes done over the whole corpus measured in WER between the raw text and the edited text. We observe that on average 31.75% of text was changed for the MT corpus. Secondly, we present the IAA numbers in terms of AWER in two evaluation rounds. In the first IAA round, the post-edited text is compared to a post-edited text made by a second pool of three annotators. The IAA of 16.87 obtained for the round 1 could be explained by the relatively high level of changes in the text and also by the difficult nature of the MT post-editing task in general. In order to measure the fluency agreement of the post-edited text, we performed a second round of IAA in which the output text of the first round was provided to the second pool of three annotators in order to measure their agreement on the correction done during the first round of annotation in terms of IAA. The low average WER of 4.92 obtained show a high agreement with the post-editing done in the first round between three annotators. The results obtained with the MT are comparable to those obtained with the L2 corpus, this can be explained by the difficult nature of both corpora and the multiple acceptable corrections for both.
6.2. Error Analysis

There will be always cases of MT post-editing disagreement, as there is often many ways to correct a given translation. With our guidelines, we try our best to reduce the inconsistency in the annotation. In Table 2, we show an example of disagreement among the annotators including a case of two acceptable corrections. For instance, Annotator C added the unnecessary word 'while' which...
was not present in the original sentence or the MT output, moreover, she kept the word ‘Uranium’ wrongly present in the MT output. On the other hand, the annotator A produced the perfect translation while the annotator B produced an acceptable one.

7. Conclusions

We have presented in detail the methodology used to create a 100K-word English to Arabic MT manually post-edited corpus, including the development of the guidelines as well as the annotation procedure and the quality control procedure using frequent inter-annotator measures. The created guidelines will be made publicly available and we look forward to distribute the post-edited corpus in a planned shared task on automatic error correction and getting feedback from the community on its usefulness as it was in the previous shared tasks we organized for the L1 and L2 corpus (Mohit et al., 2014; Rozovskaya et al., 2015).

We believe that this corpus will be valuable to advance research efforts in the machine translation area since manually annotated data is often needed by the MT community. We believe that our methodology for guideline development and annotation consistency checking can be applied in other projects and other languages as well. In the future, we plan to increase the size of the corpus and also to add other corpus domains.

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