# Rule-Based Preordering on Multiple Syntactic Levels in Statistical Machine Translation

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#### **Abstract**

We propose a novel data-driven rule-based preordering approach, which uses the information of syntax tree and word alignment to reorder the words in source sentences before decoding in a phrase-based SMT system between English and Chinese. The preorering algorithm extracts reordering patterns from multiple levels of the syntax trees of the training data and applies the rules on the source sentences in a similar manner. We've conducted experiments in English-to-Chinese and Chinese-to-English translation directions. Our results show that the approach has led to improved translation quality both when it was applied separately on the baseline or when it was combined with some other reordering approaches. We report an improvement of 0.43 in BLEU score when our preordering approach was used in addition to the short rule [1], long rule[2] and tree rule [3] based preordering approaches in the English-to-Chinese translation direction. There's also an improvement of 0.3 in BLEU score when our preordering approach was used in addition to the aforementioned preordering approaches in the Chinese-to-English translation direction. Through the translation examples, we've also found improvements in syntactic structure with our preordering approach.

## 1. Introduction

Word order is a general issue when we want to translate text from one language to the other. Different languages normally have different word orders and the difference could be very huge. Among all the languages, Chinese is one language which is very different from English, because they belong to different language families and have long period of separately development. Both languages have a Subject-Verb-Object order, but they also have a lot of differences in word order. Especially sentences in both languages can sometimes have completely different syntactic structures. The differences may involve long-distance or unstructured position changes.

Most state-of-the-art phrase-based SMT systems use language model, phrase table or decoder to adjust the word order. Or an additional reordering model is used in the loglinear model for word reordering. However, these methods may have some disadvantages, such as some don't handle long-distance reordering, some don't handle unstructured reordering and some are rather time consuming.

Encouraged by the results from the paper [1], [2] and [3], we further propose a new data-driven, rule-based preordering method, which extracts and applies reordering rules based on syntax tree. The method is called Multi-Level-Tree (MLT) reordering, which orders the constituents on multiple levels of the syntax tree all together. This preordering method rearranges the words in source language into a similar order as they are supposed to be in the target language before translation. With the appropriate word order, better translation quality can be achieved. Especially, our preordering method is very suitable for translation between language pairs like English and Chinese, which have very different word orders. Besides, the method can also be combined with the above mentioned rule-based reordering methods to achieve better translation quality.

# 2. Related Work

Word reordering is an important problem for statistical machine translation, which has long been addressed.

In a phrase-based SMT system, there are several possibilities to change the word orders. Words can be reordered during the decoding phase by setting a window, which allows the decoder to choose the next word for translation. Reordering could also be influenced by the language model, because the language model give probability of how a certain word is likely to follow. Different language model may give different probability, which further influences the decision made by log-linear model. Other ways to change the word orders include using distance based reordering models or lexicalized reordering models [4, 5]. The lexicalized reordering model reorders the phrases by using information of how the neighboring phrases change orientations.

Another way to achieve word reordering is to detach it from decoding phase and do it separately in a pre-process before decoding, in order to reduce the time for translation. This kind of preordering approaches use linguistic information to modify the word orders. Preordering can also be rule-based, which extracts different types of reordering rules by

observing reordering patterns from the training data and apply the rules to the sentences to be translated. Depends on how the rules are defined, different information may be used such as word alignments, POS tags, syntax trees, etc.

Some early approaches use manually defined reordering rules based on the linguistic information for particular languages [6, 7, 8, 9]. Later come the data-driven methods [10, 11], which learn the reordering rules automatically.

Rottman and Vogel 2007 [1] introduced the idea of extracting reordering rules from the POS tag sequences of training data and use them for reordering. Niehues and Kolss 2009 [2] went further, and developed a method for long-distance word reordering, which works good on German-English translation task due to the long-distance shift of verbs. The method extracts discontinuous reordering rules in addition to the continuous ones, which contains a place-holder to match several words and enables the word to shift cross long distance.

Afterwards, Herrmann et al. 2013 [3] introduced a new approach to reorder the words based on syntax tree, which leads to further improvements on translation quality. The algorithm takes syntactic structure of the sentences into account and extract the rules from the syntax tree by detecting the reordering of child sequences. It also has the variant based only on part of the child sequences which is suitable for language with flat syntactic structure such as German.

However, these approaches which are bases on POS tag sequences or syntax trees are mostly designed for languages like German and are not especially adapted for languages like Chinese. As Chinese has very different word orders, a reordering approach, which can further explore the syntactic structure of Chinese and utilize this information for reordering, is desirable.

The hierarchical phrase-based translation model [12] is especially suitable for Chinese translation, and provide very good translation results. It extracts hierarchical rules by using information of the syntactic structure. Phrases from different hierarchies, or so-called phrases of phrases, are reordered during the decoding.

The idea of phrases on different hierarchies has inspired us to create this preordering method based on multiple levels of the syntax tree. Besides, we also hope to detach the reordering from decoding phase and do it separately in a preprocess before decoding, in order to reduce the time for translation. This kind of preordering approaches use linguistic information to modify the word orders.

Oracle reordering has also shown values for evaluating the potential of preordering. [13] introduced the permutation distance metrics which can be used to measure reordering quality. And [14] described how we can construct permutations from the word alignment as oracle reordering.

## 3. Motivation

The word order between English and Chinese differs very significantly. For one, the words in Chinese have generally different origins as those in English, which leads to very different vocabulary and word construction. Sometimes it is very hard to find corresponding words in the other language. For example, some prepositions in Chinese have very different usage than those prepositions in English. Also the continuous writing of Chinese without spaces makes this problem more severe, since word boundaries are not always so clear in Chinese. The text needs to be segmented first before translation. A word segmentation process is used to separate the words, but the results may not always be ideal.

For the other, both languages have sometimes very different sentence structures. Thus, a word-for-word translation between English and Chinese is often unnatural or difficult to understand. Each of them has some sentence patterns that don't exist or rarely used in the other. In Chinese, a modifier is often put before the part that it modifies. While in English, it is very common that the modifier is put after the part that it modifies. Besides, English sentences with a lot of long clauses may be more suitable to translate into several Chinese sentences, because in Chinese people don't tend to use long clauses in general.

Some typical problems of word orders between English and Chinese that we've found in the data are as follows:

- Pre-modifier instead of post-modifier. In Chinese people tend to use pre-modifier rather than post-modifier.
   This involves the position change of adverbials, relative clauses and preposition phrases during translation.
- Constuction of questions.
- Special sentence constructions. For example, Bâ-construction in Chinese and sentence constructions such as there be and inverted negative sentences in English don't have correspondence in the other language in general.
- Long distance word position change.

Not all these reordering problem can be solved well by using reordering rules based on POS tag sequence or single level of syntax tree. In order to improve the reordering between English and Chinese, we need reordering method that can handle more complicated, unstructural word order change. Inspired by the ideas of reordering on syntax tree and hierarchical phrases, we created the MLT reordering, which reorders words based on multiple syntactic levels and can handle long distance word position change and complicated word position change very well.

# 4. Multi-Level-Tree Reordering

# 4.1. Reordering on Multiple Syntactic Levels

Reordering patterns are based on multiple levels of the syntax tree. Figure 1

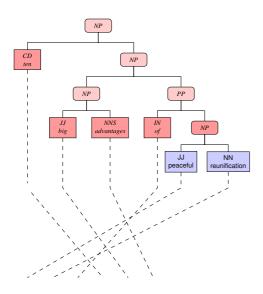


Figure 1: Reordering pattern from multiple syntactic levels

#### 4.2. Rule Extraction

#### 4.3. Rule Application

#### 5. Results

#### 6. Conclusion

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Figure 2: Schematic diagram of speech production.

# 6.5. Tables

An example of a table is shown as Table 1. Somewhat different styles are allowed according to the type and purpose of the table. The caption text may be above or below the table.

Table 1: This is an example of a table.

ratio	decibels
1/1	0
2/1	$\approx 6$
3.16	10
10/1	20
1/10	-20

## 6.6. Equations

Equations should be placed on separate lines and numbered. Examples of equations are given below. Particularly,

$$x(t) = s(f_{\omega}(t)) \tag{1}$$

where  $f_{\omega}(t)$  is a special warping function

$$f_{\omega}(t) = \frac{1}{2\pi j} \oint_C \frac{\nu^{-1k} d\nu}{(1 - \beta \nu^{-1})(\nu^{-1} - \beta)}$$
 (2)

A residue theorem states that

$$\oint_C F(z)dz = 2\pi j \sum_k Res[F(z), p_k]$$
 (3)

Applying (3) to (1), it is straightforward to see that

$$1 + 1 = \pi \tag{4}$$

Make sure to use \eqref when refering to equation numbers. Finally we have proven the secret theorem of all speech sciences (see equation (3) above). No more math is needed to show how useful the result is!

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# 7. Experiments

Please make sure to give all the necessary details regarding your experimental setting so as to ensure that your results could be reproduced by other teams.

### 8. Conclusions

This paper has described a novel approach for doing wonderful stuff such as ...

# 9. Acknowledgements

The IWSLT 2014 organizing committee would like to thank the organizing committees of INTERSPEECH 2004 for their help and for kindly providing the template files.

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