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Automatic detection of unexpected/erroneous collocations in learner corpus

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Abstract

This research investigates the collocational errors made by English learners in a learner corpus. It focuses on the extraction of unexpected collocations. A system was proposed and implemented with open source toolkit. Firstly, the collocation extraction module was evaluated by a corpus with manually annotated collocations. Secondly, a standard collocation list was collected from a corpus of native speaker. Thirdly, a list of unexpected collocations was generated by extracting candidates from a learner corpus and discarding the standard collocations on the list. The overall performance was evaluated, and possible sources of error were pointed out for future improvement.

1 Introduction

Multiword expressions (MWEs) are word combinations which present lexical, syntactic, semantic, pragmatic or statistical idiosyncrasies. The boundary between MWEs and collocations is subtle. In Ramisch et al. (2018), they defined collocations as combinations of words whose idiosyncrasy is purely statistical and show no substantial semantic idiosyncrasy. In this way they oppose MWEs to collocations. Some researchers (Sag et al., 2002) regard collocations as any statistically significant cooccurrences, which include all kinds of MWEs. Some other researchers (Garcia et al., 2019; Baldwin and Kim, 2010) consider collocations as a subset of MWEs. For Tutin (2013), collocation is a category of semantic phraseme. As defined by Mel'čuk (1998), a phraseme is a set of phrase which is not free (without freedom of selection of its signified and without freedom of combination of its components). In this sense, the meaning of phraseme is quite similar to MWE. In this research, we considered collocation as a subset of semantic phraseme and a subset of MWEs as well. To constrain the set of collocation candidates, we focus on the Verb-Noun (VN) construction.

Second language learners usually have problems with collocations. Some researchers have reported that the errors are related to the learners' L1 (Nesselhauf, 2003; Hong et al., 2011). The correction of wrong collocations¹, such as *to *create [construct] a taller and safer building*, in written essays can help learners increase their competence and thus their proficiency in English writing (Meunier and Granger, 2008). Therefore, the automatic detection and correction of erroneous collocations would be helpful for

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¹ In this research, the terms *wrong collocations*, *erroneous collocations*, *unexpected collocations*, and *collocational errors* are interchangeable.

learners. Designing such a system would support specific feedback messages that could be employed to guide learners in their meta-cognitive learning processes (Shute 2008).

Such a system may be based on two kinds of corpora: a learner corpus which is used to extract known collocational errors, and a reference corpus to extract standard English collocations (Shei and Pain, 2000). Chang et al. (2008) proposed a method of bilingual collocation extraction from a parallel corpus to provide phrasal translation memory. Their system performance was exceptionally good (precision=0.98, recall=0.91). However, this approach required a bilingual dictionary, a parallel corpus for a specific L1 and English, as well as word-alignment matching of translations.

This paper presents a preliminary research on a learner corpus. In the following sections, we will briefly explain the method, present the results, and give some discussions.

2 Method

We propose a system to extract unexpected collocations in three stages: (a) implementation and evaluation of a collocation extraction module; (b) collection of standard collocations from a native corpus; (c) extraction of wrong collocations from a learner corpus. The main principle is, firstly, to extract all possible collocations in the learner corpus, and then identify standard collocations by the reference (collocations extracted from native corpus); the remainder of the items are considered as wrong collocations. Three evaluation points were made, aiming at the collocation extraction module, the reference of standard collocations, and the extraction of wrong collocations, respectively. The system diagram and the three stages are shown in Figure 1.

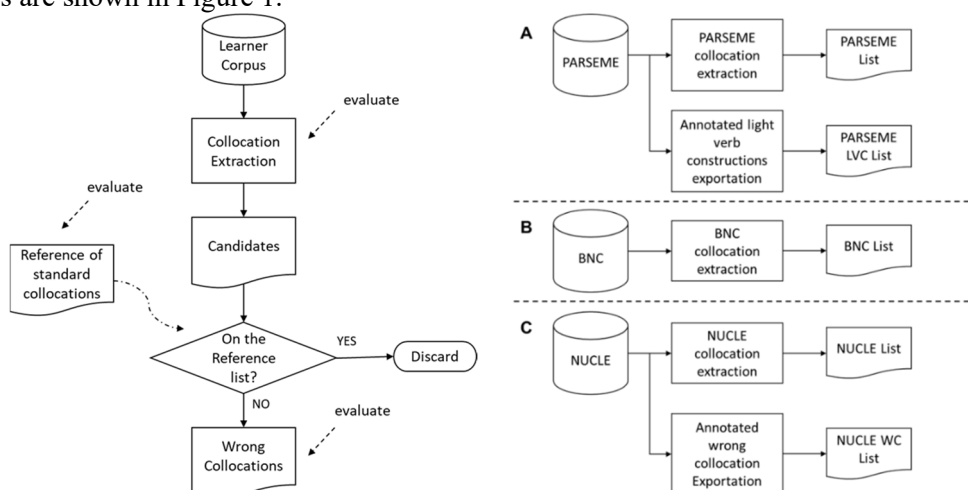


Figure 1. The system diagram and the three stages.

Stage A. Implementation and evaluation of the collocation extraction module: collocations were extracted from the PARSing and Multi-word Expressions (PARSEME²) corpus (Savary et al., 2015) with the implemented module. The results were saved as the PARSEME List. According to Garcia et al. (2019), light verb constructions (LVCs) can be regarded as collocations in VN form. The manually annotated LVCs were therefore retrieved and saved as the PARSEME LVC List. It is the gold standard (i.e. the ground truth) to evaluate the extraction module and to fine tune the parameters in the scripts.

Stage B. Collection of standard collocations: to have a large list of standard collocations, we used the implemented module to extract collocations from the British National Corpus (BNC³) (BNC Consortium, 2007) to form a list of standard collocations (the BNC List). The reference of standard collocations was built by merging the BNC List and the PARSEME LVC List. It was evaluated by manual verification. The errors in the reference list would degrade the credibility of our gold standard and thus might have a negative influence on the overall performance.

Stage C. Extraction of wrong collocations: we used the implemented module to extract candidate collocations (named as the NUCLE List) from the National University of Singapore Corpus of Learner

² <https://lindat.mff.cuni.cz/repository/xmlui/handle/11372/LRT-2842>

³ <https://ota.bodleian.ox.ac.uk/repository/xmlui/>

English (NUCLE⁴) (Dahlmeier et al., 2013). The sentences manually annotated with erroneous collocations (*Wci* tag) were also exported, and the VN terms in these sentences were detected and saved in the NUCLE WC List. It was used to evaluate the overall performance of our system.

The scripts⁵ were written in Python with Natural Language Toolkit (NLTK)⁶ (Bird and Loper, 2004). Five lexical association measures were used in collocation extraction tasks, namely the raw frequency counting, t-test, chi-square test, log likelihood ratio, and pointwise mutual information. The formulas as well as an evaluation of 84 measures can be found in Pecina (2010).

3 Results

3.1 Evaluation of the collocation extraction module

To evaluate the module, we extracted the collocations from PARSEME and compared them with the PARSEME LVC List. The precision, recall, F_1 and $F_{0.5}$ scores were used as the accuracy metric. The best precision rate is 0.11 for the bigram detection with minimal frequency of 2, using raw frequency measure, and with the top 200 collocations. Meanwhile, the best recall rate is 0.11 when both bigram and trigram detection are used, and with minimal frequency equals 2 for top 300 collocations, with the log likelihood ratio or with the raw frequency measure. the best F_1 and $F_{0.5}$ are both 0.08 for the bigram detection using raw frequency measure with a minimal frequency of 2 and with top 300 collocations. Pointwise mutual information and chi-square methods cannot give good results even without applying filters. The results obtained by t-test methods are similar to raw frequency method. The window size was set to four. Shorter or longer window lengths were tried but did not have good results, which means the words of a collocation tends to co-occur in the span of four words.

3.2 Evaluation of the BNC list

For manual verification, 200 candidates were randomly sampled from the BNC list and given to an experienced English teacher. He validated firstly obvious collocations like *take place*. For the candidates that he was not sure about, he consulted the Corpus of Contemporary American English (COCA) collocate search tool⁷. If he found the candidate in the COCA corpus, it was validated; if not, the candidate was discarded. The final precision rate is 0.57.

3.3 Intersections between lists

Ideally the union of the BNC List and the PARSEME LVC List (noted as **BNC U PARSEME LVC**) gives us the standard collocations, and NUCLE WC List gives the wrong collocations. Ideally there should be no overlapping in standard and wrong collocations. However, we found that there are intersections between the NUCLE WC List and the PARSEME LVC (11 collocations), between the NUCLE WC List and the BNC List (20 collocations), and between all three lists (4 collocations). The amount of this overlapping is therefore 27 ($20+11-4=27$), noted as **NUCLE WC \cap (BNC U PARSEME LVC)**; it is about 1.8% of the NUCLE WC List.

3.4 Optimization by selecting a threshold of Log Likelihood Ratio

Candidates were extracted from NUCLE and compared with the gold standard, i.e. the NUCLE WC List (1,471 erroneous VN collocations). Various thresholds of log likelihood ratio were tested for optimization. Figure 2(a) shows the global view of precision and recall versus different thresholds, and Figure 2(b) gives a zoom-in of threshold from zero to twelve. The highest precision is 0.5 when the threshold value is set to 430, where only two candidates are extracted. The precision and recall meet at the same level about 0.04 when the threshold is set to eight, and 1,408 candidates are extracted. The maximal

⁴ NUCLE is a collection of 1,414 essays (in a total of 1.2 million words) written by students who are non-native English speakers. It is available by submitting a license agreement via <https://www.comp.nus.edu.sg/~nlp/corpora.html>

⁵ Source codes are available online: https://github.com/jenyuli/wrong_collocation_extraction

⁶ <https://www.nltk.org/>

⁷ <https://www.english-corpora.org/coca/>

recall (0.83) is obtained by extracting all possible candidates (54,471), and the precision becomes extremely low (0.02).

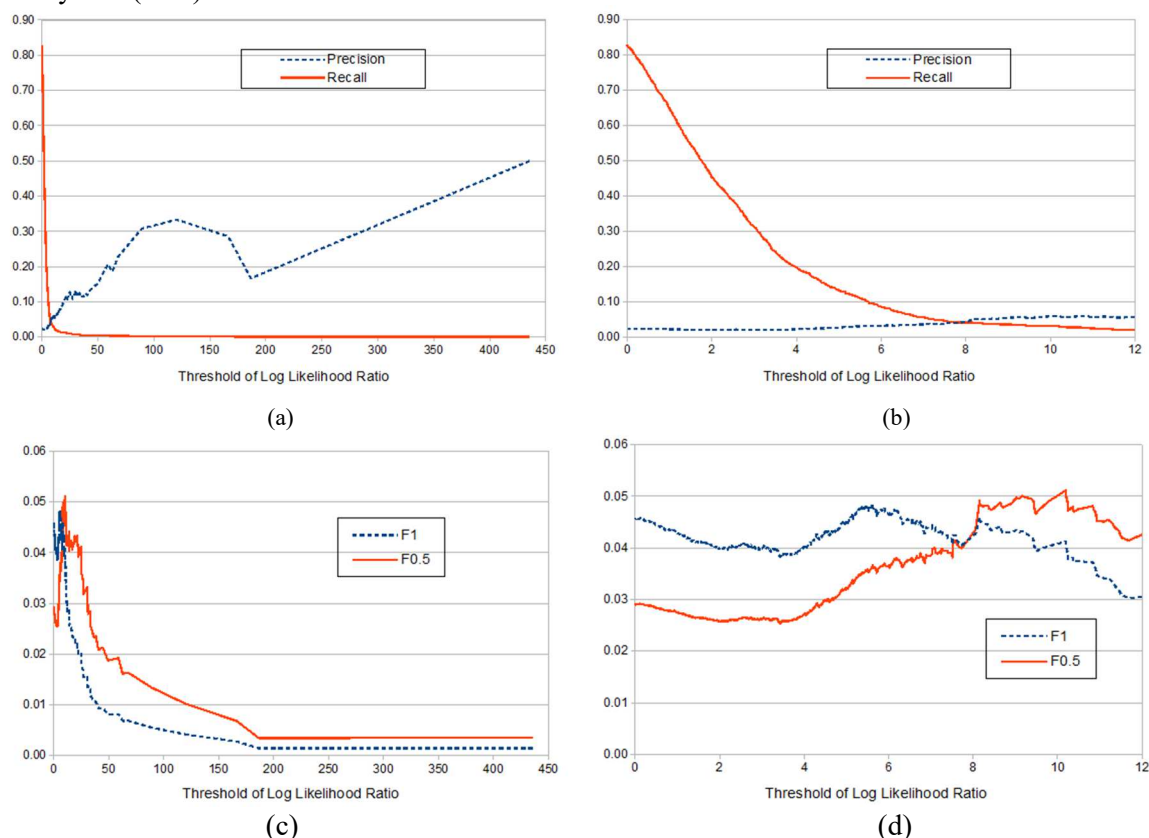


Figure 2. Precision, Recall, F₁ and F_{0.5} scores versus log likelihood ratio.

Figure 2(c) and 2(d) demonstrate the global view and zoom-in of the F₁ and F_{0.5} trends. We can see that the F_{0.5} reaches its peaks (0.05) when the threshold is set to eight or ten; while the F₁ fluctuates around 0.04 to 0.05 when threshold is set lower than eight. Considering all four indices, the optimal value of the threshold can be set about eight.

4 Discussions and conclusion

As our experiment configuration is capable to extract wrong collocations from the learner corpus, the overall performance is not satisfactory. Hence, we reviewed the results and point out some possible sources of errors for future studies.

First, regarding the PARSEME corpus, the gold standard was built based on the *LVC* tag, so it may be that the verbs of the collocations were biased. In fact, 44 out of 85 collocations on the list were constructed only by five verbs, namely *do*, *get*, *give*, *have*, and *take*. Therefore, the evaluation of the module was also biased. Regarding the BNC List, we have reached a precision of 0.57 due to the large size of corpus (100 million words) and a strict selection (top 10 for each sub-directory of the BNC). However, comparing with a previous study (Jian et al., 2004) which extracted 631,638 VN collocations from the BNC, we found that our standard collocation reference list (BNC U PARSEME LVC) was much smaller (n=942) and may have a negative influence on the performance. Regarding the NUCLE, because the Part-Of-Speech (POS) and the lemma are not available, we used a POS tagger and a Lemmatizer. Yet, their performances were not evaluated, so the gold standard NUCLE WC List was not perfectly accurate. As for the whole system, it may be helpful to incorporate a word dependency parser module to identify the object noun which received the action of the verb.

Our approach has shown a method to detect erroneous collocations in learner English. As it relies on the accurate extraction of a reference list, our next step will consist in exploring larger corpora for extraction. Such an extraction module would be of great benefit as part of a Computer Aided Language Learning System dedicated to the analysis of phraseology in learner texts.

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