Automatic Difficulty Assessment for Chinese Texts

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Abstract

We present a web-based interface that automatically assesses reading difficulty of Chinese texts. The system performs word segmentation, part-of-speech tagging and dependency parsing on the input text, and then determines the difficulty levels of the vocabulary items and grammatical constructions in the text. Furthermore, the system highlights the words and phrases that must be simplified or re-written in order to conform to the user-specified target difficulty level. Evaluation results show that the system accurately identifies the vocabulary level of 89.9% of the words, and detects grammar points at 0.79 precision and 0.83 recall.

1 Introduction

Reading is critical to foreign language acquisition (Krashen, 2005). While language textbooks provide a convenient source of reading materials, these materials are limited in quantity and do not always match the language learners’ interest. To supplement textbooks, teachers often utilize texts from other sources, such as newspapers, magazines and the web. Since they were not originally written for pedagogical purposes, these texts typically require adjustments: teachers must simplify or re-write difficult vocabulary items and grammatical constructions so that the text becomes “comprehensible input” (Krashen and Mason, 2015) to the learners; conversely, teachers might desire more advanced language usage to challenge the learners. This editing process can be time consuming and labor intensive.

To assist the editor, we built a web-based interface that automatically determines the difficulty level of Chinese texts. It detects vocabulary items and grammar points covered by the Hanyu Shuiping Kaoshi (HSK) guidelines, the official curriculum for Chinese as a foreign language (CFL) in mainland China. Furthermore, the editor can specify a target difficulty level, and ask the interface to highlight all words and grammatical constructions that must be simplified or re-written to reach the target level.

To the best of our knowledge, this is the first system that assists editors of CFL pedagogical material by explicitly pinpointing the words and grammatical constructions that exceed the target difficulty level in an official curriculum.

2 Previous Work

Most text difficulty assessment systems aim at native speakers, both for Chinese (Chen et al., 2013; Sung et al., 2015) and for other languages (Pitler and Nenkova, 2008; Sato et al., 2008). Among those that target language learners, most give a holistic score on the overall difficulty level of the text (François and Fairon, 2012; Pilán et al., 2014), but do not specifically indicate the difficult words or grammatical constructions. Hence, while these systems can help identify suitable reading material for language learners (Brown and Eskenazi, 2004), they are not designed to facilitate editing of language teaching materials, which is the goal of our system.

Targeting learners of English as a foreign language, FLAIR (Chinkina et al., 2016) can detect 87 linguistic forms in the official English curriculum in a German state. The system attains an average precision and recall of 0.94 and 0.90 in detecting grammar points. Most systems for CFL determine the difficulty level of a text on the basis of vocabulary difficulty alone. ChineseTA (Chu, 2005), for example, estimate vocabulary difficulty on the basis of word frequencies interpolated from var-
Table 1: Vocabulary and grammar difficulty level of an example sentence, according to the HSK scale. “NR” refers to a proper noun; 6+ is the vocabulary level attributed to words not found in the HSK vocabulary lists.

| LV | Vocab. items | Gram. points | LV | Vocab. items | Gram. points |
|----|--------------|--------------|----|--------------|--------------|
| 1  | 150          | 35           | 4  | 1200         | 38           |
| 2  | 150          | 58           | 5  | 2500         | 39           |
| 3  | 600          | 68           | 6  | 5000         | 28           |

Table 2: Number of vocabulary items and grammar points at each HSK level

Figure 1: Parse tree pattern, in Stanford Dependencies for Chinese, for detecting the grammar point “Determiner and classifier”

Figure 1: Parse tree pattern, in Stanford Dependencies for Chinese, for detecting the grammar point “Determiner and classifier”

3 System Description

For Chinese as a foreign language, the two major assessment scales are the Test of Chinese as a Foreign Language (Zeng, 2014) and the Hanyu Shuiping Kaoshi (HSK) (Hanban, 2014). Both contain six levels and can be mapped to the Common European Framework of Reference for Languages, a global standard for measuring foreign language proficiency. Our system adopts HSK, the more widely used of the two in mainland China.

Upon input of any Chinese passage, the system performs word segmentation, POS tagging and dependency parsing using the Stanford Parser (Manning et al., 2014). It then offers difficulty assessment in terms of vocabulary and grammar (Section 3.1), and guides the user in editing the sentence towards the target difficulty level (Section 3.2).

3.1 Difficulty assessment

The HSK guidelines provide a vocabulary list and a set of grammar points for each level; as shown in Table 2, there are a total of 9,600 vocabulary items and 266 grammar points. For vocabulary assessment, the system matches each word with these lists, but does not assess the difficulty level of proper nouns, except those included in the HSK scheme. Table 1 shows an example sentence; the vocabulary difficulty levels of its word range from level 1 (e.g., tai ‘too’) to 6+ (e.g., xiezhen ‘realism’); Qibaishi, a proper name, is not assigned any level.

Most grammar points in the HSK guidelines provide concrete examples. The only exception is the grammar point for quadrasyllabic idiomatic expressions (成語), for which we use a list of about 1,000 expressions collected from Wiktionary. Further, three grammar points — semantic passive (意...
Difficulty assessment for Chinese text

If the user specifies the target vocabulary level and grammar level, the interface highlights in yellow all words that exceed the target level, and underlines in red all words participating in grammar structures that exceed the target level. For detailed information, the user can mouse over each word to view the vocabulary level detected, as well as the name of the grammatical structure detected (Figure 2). The user can edit the text accordingly, then re-submit the updated version for assessment, in an iterative manner until the text reaches the desired level of difficulty, or when the percentage of words exceeding the level falls below an accepted threshold, as shown by the distribution of statistics at the bottom of the page.

In case the system’s word segmentation is inaccurate, the user may correct it and re-submit the text with the option “Words already separated by space”, thereby asking the system to adopt the manual segmentation.

Table 3: Statistics of the evaluation dataset

| Level | # sentences | # words | # grammar points |
|-------|-------------|---------|-----------------|
| 1     | 18          | 105     | 69              |
| 2     | 51          | 407     | 296             |
| 3     | 52          | 639     | 403             |
| 4     | 60          | 1241    | 540             |
| 5     | 65          | 1211    | 577             |
| 6     | 85          | 1970    | 801             |

4 Evaluation

In order to evaluate system performance, we harvested sentences from sample HSK exams from levels 1 to 6, obtained from the chinesetest.cn website. Our dataset contained a total of 331 sentences, including all sentences in the “Reading” sections of the examination papers for levels 1 to 4, and all sentences from reading comprehension exercises for levels 5 and 6. We performed manual word segmentation on these sentences, and annotated the HSK levels of each individual word and grammatical construction; Table 3 shows statistics of this dataset.

We evaluated system performance on both vocabulary and grammar assessment on this dataset; Table 4 presents the results according to HSK level. For vocabulary assessment, using automatic word segmentation, the system correctly recog-
Table 4: System accuracy on vocabulary assessment, and precision and recall on grammar point detection

| Level | Vocabulary Accuracy | Grammar Accuracy | Precision | Recall |
|-------|--------------------|------------------|-----------|--------|
| 1     | 0.810              |                  | 0.747     | 0.812  |
| 2     | 0.958              |                  | 0.962     | 0.865  |
| 3     | 0.890              |                  | 0.960     | 0.896  |
| 4     | 0.895              |                  | 0.649     | 0.778  |
| 5     | 0.898              |                  | 0.739     | 0.842  |
| 6     | 0.891              |                  | 0.670     | 0.777  |

The system recognized overall 89.9% of words and their vocabulary level. Most errors are due to word segmentation errors during automatic parsing, or misrecognition of proper names.

The average precision and recall of grammar points are 0.788 and 0.828. The system performs best in categories involving lexical features with unambiguous POS, such as “Pronouns” (人称代词), and worse in categories that require accurate dependency parsing, such as double object (双宾语). Errors in recall were mostly due to the non-exhaustive nature of the examples in the HSK guidelines. Precision is most challenging for grammar points that can be disambiguated only through semantic analysis, for example between the use of hui (会) to express ability vs. prediction.

5 Conclusions and future work

We have presented a web-based interface that automatically assesses the difficulty level of a Chinese text. The system indicates the vocabulary level and grammar level of specific words and grammatical structures according to the HSK scale, and highlights those that need to be simplified or re-written in order for the text to conform to the target level. We have also reported the performance of the system on vocabulary and grammar level assessment.

In future work, we plan to estimate the overall difficulty level of a sentence; to offer suggestions for lexical simplification; and to extend the scope to other linguistic features, beyond the HSK guidelines, that can help estimate the difficulty of a text.

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