

# Analyzing the Revision Logs of a Japanese Newspaper for Article Quality Assessment

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## Abstract

We address the issue of the quality of journalism and analyze daily article revision logs from a Japanese newspaper company. The revision logs contain data that can help reveal the requirements of quality journalism such as the types and number of edit operations and aspects commonly focused in revision. This study also discusses potential applications such as quality assessment and automatic article revision as our future research directions.

## 1 Introduction

Quality journalism deserves serious consideration, particularly given the disruptions of existing publishing companies and the emergence of new publishing companies, citizen journalism, and automated journalism. Although no consensus exists for the definition of quality journalism, Meyer (2009) describes several aspects that constitute quality journalism; for example, credibility, influence, accuracy, and readability. To the best of our knowledge, this is the first attempt to analyze the large-scale revision logs of professionals in the field of journalism. In this study, we explore aspects of quality journalism through analyses of the newspaper article revision logs. More specifically, we analyze the revision processes as editors refine the drafts written by reporters so that they are of publication quality.

While our attempt is still in the early stages, this paper reports the statistics of the actual revisions made by professionals and shows the usefulness of the revision logs. We also discuss the future directions of this research, for example, the potential to present feedback to reporters, extract guidelines for ‘good’ articles, and develop systems for automatic revision and sentence merging and splitting.

## 2 Analysis of revision logs

This section describes the daily activities of a newspaper company needed to publish articles and the analysis of the revision logs.

### 2.1 Flow of editing and publishing articles

A reporter drafts an article and sends it to an editor, who has over ten years’ experience as a journalist. The editor proofreads the article and forwards it to a reviewers section. The reviewers in this section fact-check the article. Finally, designers adjust the article so that they fit in the newspaper and website. In this way, a newspaper article is revised many times from the original submission.

Figure 1 compares the text from an article written by a reporter and the same text after it has been revised by an editor. The editor revises the text using insertion, deletion, and replacement. This example also shows the operations of sentence reordering and splitting.

### 2.2 Aligning original sentences with revised sentences

Revision logs present two versions of an article: the one written by a reporter (the original version) and the final version revised by an editor (the revised version). However, these logs do not provide details about the editing process used to transform the original version into the final version (e.g., individual operations of word insertion/deletion, sentence reordering). Hence, we estimate sentence alignments between the original and revised versions using the *maximum alignment* method (Kajiware and Komachi, 2016).

The accuracy, precision, recall, and F1 score were 0.995, 0.996, 0.951, and 0.957, respectively, on a dataset consisting of 50 articles in which the correct alignments were assigned by a human.<sup>1</sup>

<sup>1</sup>We chose 0.06 for word similarity threshold and 0.70 for

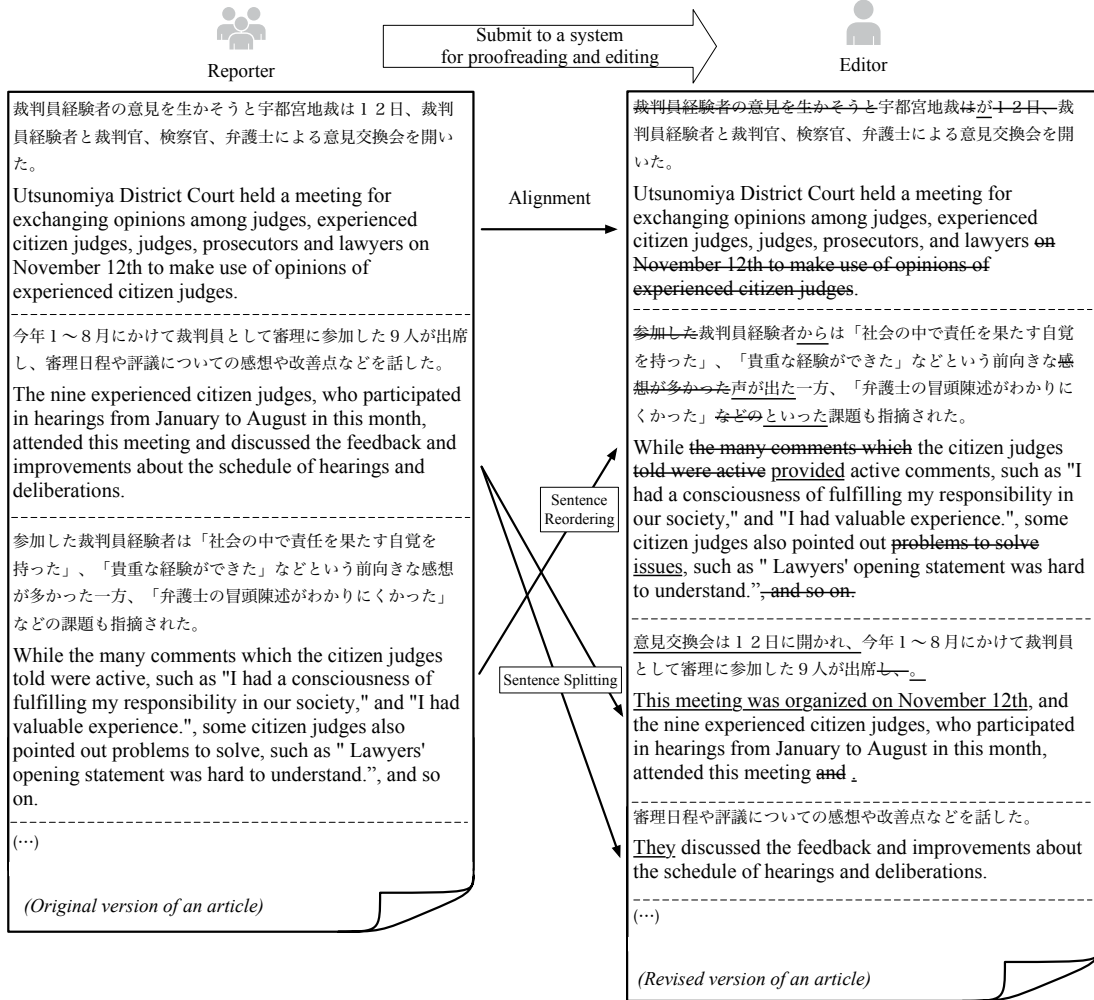


Figure 1: Comparison of the original and revised versions of some text. In the revised version, the strikethrough and underlined parts indicate deletions and insertions, respectively.

The precision, recall, and F1 score calculated from only the sentence pairs that are changed during revision were 0.926, 0.895, and 0.910, respectively. There may be some room for improving the performance of the alignments but it is sufficient for this analysis.

### 2.3 Data analysis of the revision logs

To analyze the details of the revision processes, we inspected articles published from October 1, 2015 to September 30, 2016. We applied MeCab (Kudo et al., 2004), a Japanese morphological analyzer, with the enhanced dictionary NEologd<sup>2</sup>, to split the sentences into words (morphemes) and recognize their parts-of-speech.

sentence similarity threshold, optimizing the F1 score on a development set consisting of 150 articles. We used word embeddings that are pre-trained by the original articles and revised articles to compute sentence similarity.

<sup>2</sup><https://github.com/neologd/mecab-ipadic-neologd>

The dataset analyzed in this study contains 120,331 articles with 1,903,645 original sentences and 1,884,987 revised sentences. The dataset consists of a Japanese newspaper's articles<sup>3</sup>, which have a mixed domain (genre) of the news, and most of the articles have the same writing style. We obtained 2,197,739 sentence pairs using the alignment method described in Section 2.2. The number of aligned pairs is larger than that of the sentences because an original sentence can be aligned to multiple sentences in the revised version. About half of the sentence pairs (1,108,750) were unchanged during the revision process, and the remaining pairs (1,088,989) were changed. In this section, we report the statistics of the edit operations in the changed pairs. We found that newspaper companies produce a huge number of sentences, about half of which are revised, for analy-

<sup>3</sup>The Asahi Shimbun Company provided this dataset.

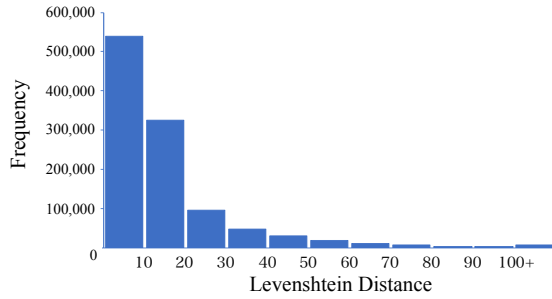


Figure 2: Distribution of the Levenshtein distance of changed sentence pairs.

Original sentence:

市場に同じ魚が出回りすぎると、魚の単価が下がってしまう。

If the same kind of fish is distributed in large quantities in the market, the unit price of the fish manages to decrease.

Revised sentence:

市場に同じ魚が出回りすぎるの水揚げが重なると、魚の単価が下がってしまう。

If the same kind of fish is distributed landed in large quantities in the market, the unit price of the fish manages to decrease is decreasing.

Figure 3: An example of the original and revised sentence pair whose Levenshtein distance is 15.

sis just within a year.

Figure 2 presents the distribution of the Levenshtein distance between the original and revised sentences. The mean of the Levenshtein distances of the revised pairs (15.04) indicates that the dataset includes many examples in which drafts are deeply edited by the editors. Figure 3 is an example of the sentence pair which has the mean of the Levenshtein distance of the dataset (15).

Table 1 lists the number of insertions, deletions, and replacements, according to the number of words involved in the edit operations. We found that 56.20% of the total edit operations were replacements for one or two words, and this fact indicates that editors revised these articles with impressive attention to detail.

Table 2 shows the number of edit operations separated by different part-of-speech. The most frequent target for revisions is nouns, followed by particles (postpositions). This result indicates that revisions in terms of both content and readability are important for improving the quality of articles.

# of words	Insertion	Deletion	Replacement
1	139,790	160,975	1,424,118
2	118,261	151,641	303,293
3	57,397	53,789	115,525
4	35,272	31,719	75,909
5	21,339	20,435	33,805
6	13,295	14,756	21,419
7	14,599	13,030	24,400
8	8,631	9,301	10,707
9	10,196	10,760	8,475
Over 10	48,754	60,523	61,387
<b>Total</b>	<b>467,534</b>	<b>526,929</b>	<b>2,079,038</b>

Table 1: Number of edit operations with respect to the number of words involved.

Tag	Count
Noun	1,255,113
Noun + Noun	174,306
Particle	157,840
Symbol	128,584
Noun + Particle	106,548
Verb	85,709
Symbol + Noun	47,635
Particle + Noun	42,714
Particle + Verb	41,342
Prefix	41,194
Noun + Symbol	37,580
Verb + Auxiliary	20,836
Auxiliary	18,145
Noun + Verb	14,153
Adverb	9,009
Others	101,714

Table 2: Distribution of parts-of-speech as targets for the edit operations involving one or two words.

### 3 Future directions for quality assessment and automatic article revision

There are several possible future directions for the utilization of the revision logs.

#### 3.1 Feedbacks to reporters

We can use the revision logs for improving the writing skills of reporters. An interesting finding in the revision logs is that the articles of young reporters (1–3 years’ experience) tend to be revised more than those of experienced reporters (31–33 years’ experience): the mean Levenshtein distances of these young and experienced reporters are 15.82 and 12.95, respectively. As exemplified by this finding, the revision logs can indicate the main types of revisions that a particular group of reporters or an individual reporter receives. We will explore the potential of the revision logs for assessing the writing quality of a reporter and presenting them with feedback.

### 3.2 Establishing guidelines for writing articles

Most textbooks on Japanese writing (including the internal handbook for reporters produced by the newspaper company) recommend that a Japanese sentence should be 40 to 50 characters long (Ishioka and Kameda, 2006). We could confirm that the newspaper articles satisfy this criterion: the revised sentences are 41.10 characters long on average. In this way, we can analyze the revision logs to extract various criteria for establishing the guidelines for ‘good’ articles.

### 3.3 Automatic article revision within sentences

Another future direction is to build a corpus for improving the quality of articles. The revision logs collected for a year (excluding duplicates) provide 517,545 instances of replace operations, 79,639 instances of insertions, and 54,111 instances of deletions that involve one or two words. Table 3 shows some instances of the replace operations. It may not be straightforward to use the revision logs for error correction because some edit operations add new information and remove useless information. Nevertheless, the logs record the daily activities of how drafts are improved by the editors. In future, we plan to build an editing system that detects errors and suggests wording while the reporters write drafts. We can use natural language processing techniques for these tasks because local error correction has been previously researched (Cucerzan and Brill, 2004).

### 3.4 Automatic sentence merging and splitting

The alignment method found 69,891 instances of sentence splitting (wherein an original sentence is split into multiple sentences) and 68,550 instances of sentence merging (wherein multiple original sentences are merged into one sentence). Table 4 shows examples of sentence splitting and merging. We observe some sentences are also compressed during sentence merging and splitting. We can use these instances as a training data for building a model for sentence splitting and merging (with compression), and this may be an interesting task in the field of natural language processing.

## 4 Conclusion

In this paper, we explored the potential of the revision logs of a newspaper company for assessing

Original	Revised
同政府関係者 this Government officials	韓国政府関係者 Korean Government officials: specification
放射線汚染 contamination by radial ray	放射能汚染 radiologically contamination
破顔し broke into a smile	笑顔で話し spoke with a smile: simplification
バラティ Varety	バラエティー Variety: typo
タンパク質 Protain: written in Katakana and Kanji	たんぱく質 Protain: written in Hiragana and Kanji
買えた could buy	買える can buy

Table 3: Examples of commonly replaced words/phrases.

the quality of articles. In addition to presenting the revision logs statistics, we discussed the future directions of this work, which include feedback to reporters, guidelines for ‘good’ articles, automatic article revision, and automatic sentence merging and splitting.

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Splitting	
(S1)	<p>新たな窓口を設けるなど内部通報制度も強化し、通報は問題が発覚する前の 88 件 (14 年度) から 263 件 (15 年度) と 3 倍に増えた。</p> <p>They enhance whistle-blowing systems by providing such as new counseling offices, and the number of whistle-blowing was increased three times from 88 in 2014, in which this issue was found out, to 263 in 2015.</p>
(S2)	<p>新たな窓口を設けるなど内部通報制度も強化。通報は 2015 年度に 263 件と、不正会計問題の発覚前の 14 年度の 88 件から約 3 倍に増えたという。</p> <p>They enhance whistle-blowing systems by providing such as new counseling offices. As a result, the number of whistle-blowing was increased three times from 88 in 2014, in which this issue was found out, to 263 in 2015.</p>
Merging	
(M1)	<p>同署によると、事務所南側 1 階の窓が割られ、室内にある防犯カメラのモニター 4 台がすべて壊されていた。食器棚も倒され、食器が散乱していたという。</p> <p>Police said that the window on the first floor of the office south is broken, and all four displays for the security camera was destroyed. Police also said that the cupboard was knocked down, and the dished are scattered in the room.</p>
(M2)	<p>署によると、室内の防犯カメラのモニター全 4 台が壊され、食器棚が倒れて食器が散乱していた。</p> <p>Police said the all four displays for the security camera was destroyed, the cupboard was knocked down, and the dished are scattered in the room.</p>

Table 4: Examples of sentence splitting and merging. Sentences S1 and M1 are the original sentences, and S2 and M2 are the revised sentences. In the merging example, we can also observe the sentence compressing; the part “the window on the first floor of the office south is broken” was eliminated in M2.