7LN003 Machine Translation

Presentation Essay

An Evaluation on Google and Weblio
Online Machine Translation Services
in Japanese to English Translation

FONG HEI MAN, HALEY

STUDENT NUMBER: 1913071

May 2020

1 In	troduction	3
1.1	Machine Translation (MT) and Online MT Services	3
1.2	MT Evaluation	3
1.3	Challenges in Japanese to English Translation	4
2 M	lethodology	5
2.1	Experimental Procedures	5
2.1	.1 Text Collection	5
2.1	2 Text Preprocessing and Alignment	5
2.1	Translation with Online MT Services	5
	2.1.3.1 Google Translate (https://translate.google.com/)	5
2.1	2.1.3.2 Weblio (https://translate.weblio.jp/) .4 Evaluation of Translation Output	6
2.2	Experimental Setup	7
2.2	Experimental Secup	,
3 R	esults and Discussion	7
3.1	Experiment 1: Children's Fiction in full Japanese characters	8
3.2	Experiment 2: Children's Fiction rewritten with Chinese characters	9
3.3	Experiment 3: News Article	10
4 D	iscussion	11
5 C	onclusion	11
6 R	eferences	12

1 Introduction

1.1 Machine Translation (MT) and Online MT Services

Machine translation (MT) refers to the use of computerised systems to produce translations from a natural language to another (Hutchins and Somers, 1992). The process can either be or be not interfered by humans. Depending on the extent of human interference, types of MT can be classified across a spectrum -- from Human Translation, Computer-Aided (Human) Translation (CAT), Human-Aided Machine Translation (HAMT), and finally the ideal of Fully-Automated High-Quality Translation (FAHQT).

Over the history of MT, various approaches towards achieving the ideal goal of FAHQT has been put forward. Rule-based approaches were one of the earliest steps taken in MT, although not being welcomed for a period, it made a comeback in the 1980s. Along with advanced computers, new approaches such as example-based MT and statistical-based MT emerged. Since the 1990s, the development of MT has been speeding up exponentially, thanks to the rapid development of the Internet. Not only is translation made easier and more accessible to people, innovative approaches to MT using machine learning and neural networks further closed the distance left towards FAHQT. Nowadays, everyone can be a translator as long as they are connected to the Internet and have access to relevant websites.

This essay looks into the performances of two online translation services, *Google* and *Weblio*, in Japanese to English translation. By translating texts that are highly available to and frequently accessed by laymen, the quality of the services would be assessed to answer the following questions: *Whether the translation services are good enough for daily use by ordinary people* and *whether the systems are reliable for translating languages of specific purposes (LSPs)*. In other words, how useful are the system in professional fields.

1.2 MT Evaluation

With numerous options in the market, various MT evaluation proposals have been put forward in the translation community. The performance of an MT system is important to a few stakeholders in MT -- the developers, the translators and the audience of the translation (Hutchins and Somers, 1992). For instance, with MT evaluation, developers can optimize translation quality and efficiency of systems, reducing the workload of human translators (in either pre- or post-editing), and ultimately providing natural, high quality translations to message recipients.

MT evaluation can be performed either by computer programmes or humans. For automatic MT evaluation, the programme takes in the MT output and reference translation, then returns a score showing the similarity of the two. Most automatic MT evaluating methods score translations using string-based methods, comparing the inputs in units of characters, n-grams or sentences. Common scoring methods include BLEU (Bilingual Evaluation Understudy), METEOR (Metric for Evaluation of Translation with Explicit Word Ordering) and the Levenshtein (Edit) Distance. For human MT evaluation, since human perception is subjective, frameworks have been proposed to guide human evaluators to assess the quality of MT outputs. Adequacy, fluency and error identification are the main aspects being considered when rating an MT output.

1.3 Challenges in Japanese to English Translation

One of the largest challenges in Japanese-English translation is to preserve sociolinguistic information. In Japanese, sociolinguistic information is encoded into either different terms having the same meaning, or in the suffixes of verbs phrases. Meanwhile, orthography is one of the obstacles in meaning disambiguation. In the Japanese writing system, *kanas* and *kanjis* are used. The former are Japanese-invented characters, in which one character represents one syllable. The latter are Chinese characters that are either imported from China or invented locally in Japan. The same combination of *kanas* can result in different *kanjis* while one *kanji* can have multiple pronunciations, meaning that it can be broken down into different combinations of *kanas*. Last but not least, ellipsis is common in Japanese discourse (Okamoto, 1985). Noun phrases and verb phrases are often left out as a result of the distinctive communication style of a high-context culture. Information like the referents of pronouns and the performers or recipients of actions are omitted, and it is expected that listeners or readers can pick up such information from the surrounding context. The table below demonstrates the summary of these difficulties for the sentence *l am Hanako*.

	Casual	Watashi-wa / Hanako – da.
Formality	Formal	Watashi-wa / Hanako – desu.
Torritainty	Honorific	Watakushi -wa / Hanako – degozaimasu.
	Dialect	Watashi-wa / Hanako – ja.
	Kana	わたし‐は/はなこ/です。
Orthography	(Possible Kanjis)	(わたし → 私 or 渡し / はなこ → 華子 or 花湖)
or mography	Kanji	私-は/花子/です。
	(Possible Kanas)	(私 → わたし or わたくし or し / 花子 → はなこ or かし)
Ellipsis	Omitting pronoun	(Watashi-wa /) Hanako – desu.

2 Methodology

2.1 Experimental Procedures

2.1.1 Text Collection

Two pieces of texts were taken for the study – a children's fiction and a news article. The texts were around 1500 and 1700 characters-long respectively.

For the piece of children's fiction, it was retrieved from an online collection of bilingual stories¹. The story retrieved was written in full Japanese characters, and each line of the story was aligned with an English translation.

The news article is was retrieved from BBC. The article was originally written in English² and a translated version is available on the Japanese version of the website³. Hyperlinks to the articles are supplied at the bottom of the page.

2.1.2 Text Preprocessing and Alignment

A third piece of source text was created by re-writing the story from Japanese characters into Chinese characters where appropriate. This was to test whether the use of Chinese characters facilitate meaning disambiguation during the translation process. The resulting document consisted of around 1300 characters.

Moving onto sentence alignment, each line of the story was aligned with the respective translation into an excel file for comparison. For the news article, sentences that only exist in either the English or the Japanese version were removed so that a parallel comparison could be made. Only the sentences that were remained would be passed into the MT systems in the next stage.

2.1.3 Translation with Online MT Services

2.1.3.1 Google Translate (https://translate.google.com/)

Google began providing online translation service since 2006. (Turovsky, 2016) It started off by using statistical MT, with training data from corpora of texts from the United Nations and the European Union. It also invites its users, the Translation Community to make contributions to improve their MT system. For instance, users are invited to provide their own translations to phrases provided by the system. Or, they can help rate possible translation so that the system learns which ones are better. In recent years, Google has upgraded their MT service by using deep learning (Turovsky, 2017).

¹ Yukionna (The Snow Woman), from Hukumusume Fairytale Collection http://hukumusume.com/douwa/English/jap/01/22 j&E.html

² Coronavirus: World Health Organization members agree response probe https://www.bbc.com/news/world-52726017

³ WHO、新型ウイルス対応を検証へ 全加盟国が同意 <u>https://www.bbc.com/japanese/52733876</u>

2.1.3.2 Weblio (https://translate.weblio.jp/)

Weblio is a Japanese company that has been providing online dictionary services since 2005. Their online translation services has been available since 2012, offering Japanese to/from Chinese, Korean and English translation. Weblio provides its users with support on studying foreign languages (mainly English) for overseas employment and travel (Weblio, 2020). For example, there is a sample-sentence-lookup function on their website. Users can input a Japanese sentence and the system returns English translations under different contexts for users to reference. Examples are from their self-developed email corpus and the Wikipedia Japanese-English parallel corpus. While the approach of MT is not specified on Weblio's website, other services provided by the website suggested that statistical MT is employed.

2.1.4 Evaluation of Translation Output

The translation outputs from the online systems were evaluated both automatically and manually.

BLEU was used as the automatic evaluation method. BLEU stands for "BiLingual Evaluation Study", in which a numerical metric is utilized to measure the closeness between a MT output with a gold, corpus standard (Kishore, Roukos, Ward and Zhu, 2002). The BLEU scores for each sentence in the output was calculated. The score of a piece of output would then be computed by averaging the sum of the scores of each sentence. The more the BLEU score is closer to 1, the better the quality is the MT output. Copying the MT outputs into a .txt file, BLEU score for the outputs were generated using the Natural Language Toolkit in Python. Here are the source codes of the evaluation program (Brownlee, 2017).

```
1 from nltk.translate.bleu_score import sentence_bleu
2 ### Prepairing the standard
3 text = open("standard_YUKI.txt")
4 standard = []
5 for sent in text:
6 sentence = sent.split()
   #print(sentence)
    if sentence != []:
    standard.append(sentence)
9
10 #print(standard)
1 ### Preparing the output
2 fn = "google_YUKI_1.txt'
3 translation = open(fn)
4 output = []
5 score = 0
6 \text{ count} = 0
7 for trans in translation:
8 t = trans.split()
   #print(trans.split())
10 #output.append(t)
11
    if t != []:
     #print(t)
12
13
     count += 1
14
     score += sentence bleu(standard, trans.split())
15 #print(len(output))
16 print("Average BLEU score for *", fn, "*:", score/count)
```

As BLEU only shows the similarity of the MT output with a reference, it could only tell how accurate a result is with respect to the reference. Whether the translation is readable or appropriate in the target culture is yet to be judged (Hutchins and Somers, 1992). This is why manual evaluation is necessary, and hence the results would be evaluated by a human, who would comment on the outputs' accuracy and fluency.

2.2 Experimental Setup

Three trials of experiments were carried out to access the performance of the MT system. The first using the piece of children's fiction written in full Japanese characters and the second with the one written with Chinese characters. These trials assessed the importance of meaning disambiguation using Chinese characters, as well as the ability of the systems to translate everyday expressions such as onomatopoeic words and dialectal expressions, which do not have direct English translations. Last but not least, the last trial was performed by back-translating the news article and see how well the systems performed with texts of a more formal and informative register.

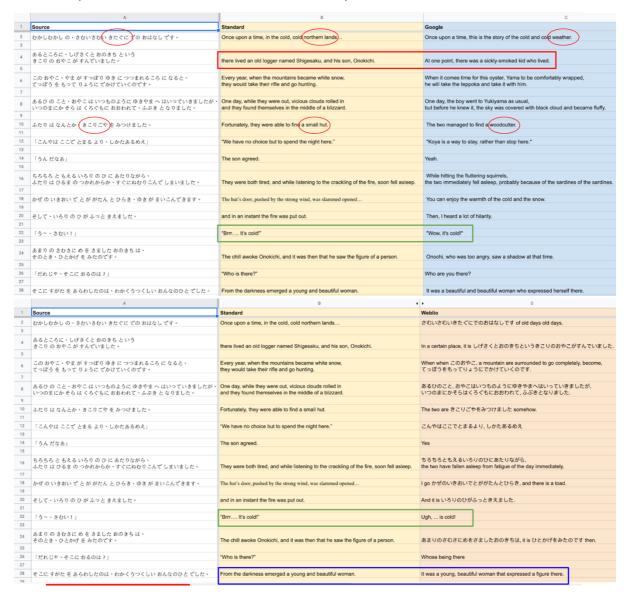
3 Results and Discussion

BLEU scores for the experiments are listed as below:

	Google Translate	Weblio
Story (in full Japanese characters)	0.5865	0.5469
Story (with Chinese characters)	0.6160	0.6134
News article	0.4339	0.4779

Comments on the systems would be made in this section, extracting examples from the translation results for illustration. The full results in a spreadsheet format could be accessed from https://docs.google.com/spreadsheets/d/1KUr-tcNDI5tvH4hkaF7P-UibVPFbM93-lveQOM_FNro/edit?usp=sharing .

3.1 Experiment 1: Children's Fiction in full Japanese characters



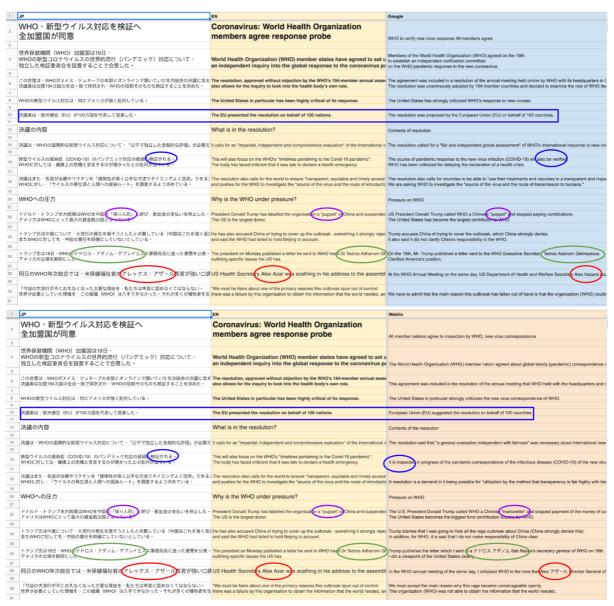
In this experiment, Google performed significantly better than Weblio as reflected from the BLEU scores. Weblio's performance was in fact very poor as only 19 out of 52 lines of the story was fully translated into English. A large amount of instances remained in Japanese in the output text. Yet, despite of Google's effort to translate the whole text into English, a number of instances were incorrectly translated due to ambiguity with the Japanese characters (marked in red). As a result, even if Google's results were readable, they did not make much sense. Besides, it was noticed that the tone and emotion of characters' speech was incorrectly translated. However, even if Weblio made few attempts to fully translate sentences into English, it did correctly captures the emotions (as marked in green). In sentences that Weblio was able to translate, the quality of the results were fair. Terms were correctly translated but it was obvious that better candidates are available (marked in blue).

3.2 Experiment 2: Children's Fiction rewritten with Chinese characters



From the BLEU scores, it is well-acknowledged that Chinese characters are helpful in meaning disambiguation, especially for Weblio where a jump in the score was observed. Reviewing Google's errors in the first experiments, the nouns and names were correctly identified in this experiments as circled in green. Still, it failed to capture characters' emotion. Meanwhile, having a better look at Weblio's translation, it is fair to say that when Weblio had no problems in correctly translating nouns and using accurate candidates (annotated in green), but it continued to show incapability in translating dialectal and onomatopoeic expressions (circled in blue), which was never a problem for Google. On the other hand, Weblio inserted a pronoun in one of the narrations while it was not in the original text.

3.3 Experiment 3: News Article



It was unexpected that Weblio outperformed Google from the BLEU scores, especially when it once again gave up translating when it came across a relatively new concept. Circled in red and green are the names of current officials, which were translated in Google's output but not in Weblio's. This suggested that Google's neural MT system is capable of more frequent updates with help from its Translate Community, and that Weblio's fails to catch up with the fast-changing world. Comparing Google's and Weblio's outputs, the translation of the word "puppet" sparked an interesting observation as circled in purple. Literally translating the Japanese translation of "puppet", it would be "controlled doll" as in the left column. In back-translation, the word became "marionette", which is the French word borrowed into Japanese to refer to controlled dolls. This prompted the idea that Weblio is more faithful to Japanese compared to Google, and the claim can be supported by the retention of active and passive voices in the Weblio outputs marked in blue.

4 Discussion

In terms of accuracy, while both Google and Weblio results were satisfactory, the latter has a larger room for improvement. For instance, the utilization of a user community greatly helped Google to conquer dialectal and onomatopoeic expressions when translating the story from Japanese to English. It is also noted that Weblio's translation is more faithful to the source texts, choosing candidates that are more frequent with Japanese English norms. In addition, overall speaking, the Weblio translations of the texts sounded more formal compared to Google's. This could be explained since Weblio complies data to help its users improve their English for education and work. So that formality of the data would be different to that of casual translations collected from Google's translator community where laymen are engaged as well. On the contrary, Google's outputs were of a more natural and fluent tone of English.

Referring to the first two experiments, it was proved that Chinese characters help to improve meaning disambiguation in the translation process. In fact, a counter example was found in the outputs indicating that Japanese characters, signalling the pronunciation of the Chinese characters, can be useful for meaning disambiguation too. For instance, \bot can be pronounced as fu-ta-ri (\bot)) when referring to two people, but it can be read as ni-ni-n ($(\bigcup \bigcup \bigcup)$) while counting two persons. While both systems correctly translated the term in the first experiment, Weblio failed to do so in the second experiment. This could further prove that Weblio is developed using more texts of a formal register and complex topics since Chinese characters are harder to learn and thus are used only for more sophisticated concepts.

Another point to note is that, although not fully presented above, the insertion of pronouns in narratives in the outputs. Even if Google's neural MT seemed to be winning, it occasionally fails to translate ellipsis properly. It could be concluded that Google is still weak in capturing pragmatic meanings for high-context cultures like Japanese, along with the evidence of imperfect translation of characters' speech in the first experiment.

5 Conclusion

All in all, Google Translate and Weblio Translate are easy-to-use online MT system which caters to the needs of commoners. Google is optimal for daily uses, producing natural outputs and with an excellent grasp on up-to-date information. For Japanese users, it is undoubted that Weblio serves its purpose by providing useful translations at school and at the workplace, but it is reckoned that they utilize Google for more natural English expressions. Meanwhile, with better knowledge in Japanese pragmatics, it is possible for Japanese learners to have a better understanding in the culture by studying the Japanese-style English translation as well. At the end of the day, both systems are successful in fulfilling the goals of everyday cross-border communication but could be limiting for LSP.

6 References

Brownlee, J., 2017. A Gentle Introduction To Calculating The BLEU Score For Text In Python. [online] Machine Learning Mastery. Available at: https://machinelearningmastery.com/calculate-bleu-score-for-text-python/.

Hutchins, W. and Somers, H., 1992. An Introduction To Machine Translation. London: Academic, pp.2-9, 161-174.

Papineni, K., Roukos, S., Ward, T. and Zhu, W.J., 2002, July. BLEU: a method for automatic evaluation of machine translation. In Proceedings of the 40th annual meeting on association for computational linguistics (pp. 311-318). Association for Computational Linguistics.

Okamoto, S., 1985. Ellipsis In Japanese Discourse. [ebook] Available at: https://escholarship.org/content/qt4zx1c0rg/qt4zx1c0rg.pdf.

Turovsky, B., 2016. Ten Years Of Google Translate. [online] Google. Available at: https://www.blog.google/products/translate/ten-years-of-google-translate/>.

Turovsky, B., 2017. Higher Quality Neural Translations For A Bunch More Languages. [online] Google. Available at: https://blog.google/products/translate/higher-quality-neural-translations-bunch-more-languages/.

Weblio, I., 2020. サービス内容 | Weblio, Inc.. [online] Weblio, Inc. Available at: https://www.weblio-inc.jp/service/>.