

Quality-Sensitive Test Set Selection for a Speech Translation System

Fumiaki Sugaya¹, Keiji Yasuda², Toshiyuki Takezawa and Seiichi Yamamoto

ATR Spoken Language Translation Research Laboratories

2-2-2 Hikari-dai Seika-cho, Soraku-gun, Kyoto, 619-0288, Japan

{fumiaki.sugaya, keiji.yasuda, toshiyaiki.takezawa,
seiichi.yamamoto}@atr.co.jp

Abstract

We propose a test set selection method to sensitively evaluate the performance of a speech translation system. The proposed method chooses the most sensitive test sentences by removing insensitive sentences iteratively. Experiments are conducted on the ATR-MATRIX speech translation system, developed at ATR Interpreting Telecommunications Research Laboratories. The results show the effectiveness of the proposed method. According to the results, the proposed method can reduce the test set size to less than 40% of the original size while improving evaluation reliability.

1 Introduction

The translation paired comparison method precisely measures the capability of a speech translation system. In this method, native speakers compare a system's translation and the translations, made by examinees who have various TOEIC scores. The method requires two human costs: the data collection of examinees' translations and the comparison by native speakers. In this paper, we propose a test set size reduction method that reduces the number of test set utterances. The method chooses the most sensitive test utterances by removing the most insensitive utterances iteratively.

In section 2, the translation paired comparison method is described. Section 3 explains the proposed method. In section 4, evaluation results for ATR-MATRIX are shown. Section 5 discusses the experimental results. In section 6, we state our conclusions.

2 Translation paired comparison method

The translation paired comparison method (Sugaya, 2000) is an effective evaluation method for precisely measuring the capability of a speech translation system. In this section, a description of the method is given.

2.1 Methodology of the translation paired comparison method

Figure 1 shows a diagram of the translation paired comparison method in the case of Japanese to English translation. The Japanese native-speaking examinees are asked to listen to Japanese text and provide an English translation on paper. The Japanese text is spoken twice within one minute, with a pause in-between. To measure the English capability of the Japanese native speakers, the TOEIC score is used. The examinees are requested to present an official TOEIC score certificate showing that they have taken the test within the past six months. A questionnaire is given to them and the results show that the answer time is moderately difficult for the examinees.

The test text is the SLTA1 test set, which consists of 330 utterances in 23 conversations from a bilingual travel conversation database (Morimoto, 1994; Takezawa, 1999). The SLTA1 test set is

¹Current affiliation: KDDI R&D Laboratories. Also at Graduate School of Science and Technology, Kobe University.

²Also at Graduate School of Engineering, Doshisha University.

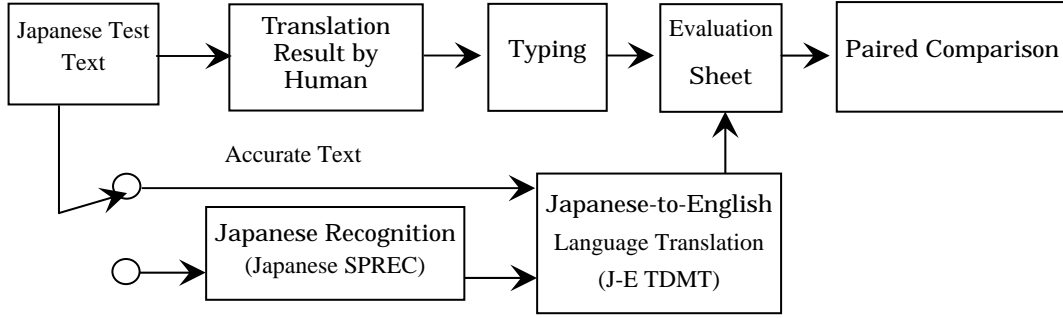


Figure 1: Diagram of translation pair comparison method

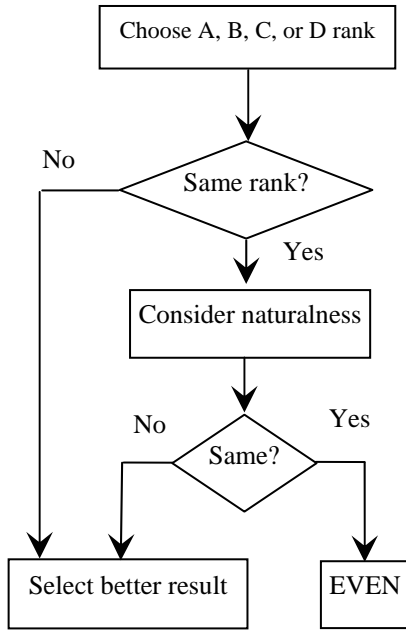


Figure 2: Procedure of comparison by native speaker

open for both speech recognition and language translation. The answers written on paper are typed. In the proposed method, the typed translations made by the examinees and the outputs of the system are merged into evaluation sheets and are then compared by an evaluator who is a native English speaker. Each utterance information is shown on the evaluation sheets as the Japanese test text and the two translation results, i.e., translations by an examinee and by the system. The two translations are presented in random order to eliminate bias by the evaluator. The evaluator is asked to follow the procedure illustrated in Figure 2. The four ranks in Figure 2 are the same as those used in Sumita (1999). The ranks A, B, C, and D indicate: (A) Perfect: no problems in both

information and grammar; (B) Fair: easy-to-understand with some unimportant information missing or flawed grammar; (C) Acceptable: broken but understandable with effort; (D) Nonsense: important information has been translated incorrectly.

2.2 Evaluation result using the translation paired comparison method

Figure 3 shows the result of a comparison between a language translation subsystem (TDMT) and the examinees. The input for TDMT included accurate transcriptions. The total number of examinees was thirty, with five people having scores in every hundred-point TOEIC range between the 300s and 800s. In Figure 3, the horizontal axis represents the TOEIC score and the vertical axis the system winning rate (SWR) given by following equation:

$$SWR = \frac{N_{TDMT} + 0.5 \times N_{EVEN}}{N_{TOTAL}} \quad (1)$$

where N_{TOTAL} denotes the total number of utterances in the test set, N_{TDMT} represents the number of "TDMT won" utterances, and N_{EVEN} , indicates the number of even (non-winner) utterances, i.e., no difference between the results of the TDMT and humans. The SWR ranges from 0 to 1.0, signifying the degree of capability of the MT system relative to that of the examinee. An SWR of 0.5 means that the TDMT has the same capability as the human examinee.

Figure 3 shows that the SWR of TDMT is greater than 0.5 at TOEIC scores of around 300 and 400, i.e., the TDMT system wins over humans with TOEIC scores of 300 and 400. Examinees, in contrast, win at scores of around 800. The capability balanced area is around a score of 600 to

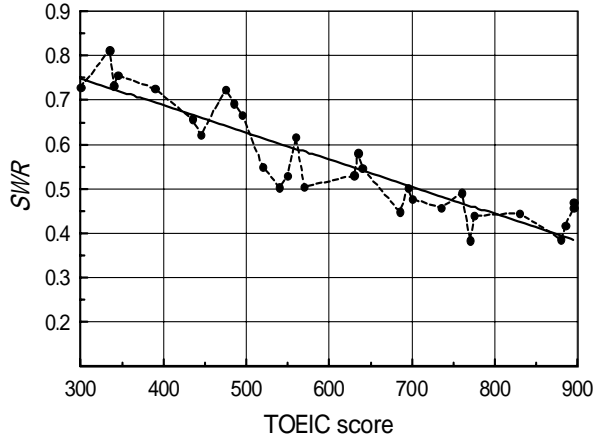


Figure 3: Evaluation results using translation paired comparison method

700. To precisely determine the balanced point, we used regression analysis. The straight line in Figure 3 is the regression line. The capability balanced point between the TDMT subsystem and the examinees is 0.5 of SWR. In Figure 3, the exact point is a TOEIC score of 708. We call this point the system's TOEIC score. Consequently, the translation capability of the language translation system equals that of the examinees at around a score of 700 points on the TOEIC.

The experimental result for ATR-MATRIX, which consists of a speech recognition subsystem and TDMT, has been also reported (Sugaya, 2000). This system's TOEIC score is 548, where the number of speech recognition errors is a factor in the degradation of the score.

2.3 Error in the system's TOEIC score

The SWR (Y_i) and TOEIC scores for the examinees (X_i) are assumed to satisfy the population regression equation:

$$Y_i = \beta_1 + \beta_2 X_i + \varepsilon_i \quad (i = 1, 2, \dots, n) \quad (2)$$

where β_1 and β_2 are population regression coefficients. The error term (ε_i) is assumed to satisfy the following condition:

- (a) $E(\varepsilon_i) = 0$
- (b) $V(\varepsilon_i^2) = \sigma^2, \quad i = 1, 2, \dots, n$ (3)
- (c) $Cov(\varepsilon_i, \varepsilon_j) = E(\varepsilon_i, \varepsilon_j) = 0$ if $i \neq j$
- (d) $\varepsilon_i \cong 0$

Under the above condition, the standard deviation of the system's TOEIC score is calculated by

$$\sigma_t = \left| \frac{\sigma}{\beta_2} \right| \sqrt{\frac{1}{n} + \frac{(C_0 - \bar{X})^2}{\sum (X_i - \bar{X})^2}} \quad (4)$$

where n is the number of examinees, C_0 is the system's TOEIC score, and \bar{X} is the average of the examinees' TOEIC scores. Equation (4) indicates that the minimum error is given when the system's TOEIC score equals the average of the examinees' TOEIC scores.

By using a t -distribution, the confidence interval (CI) of the system's TOEIC score with confidence coefficient $1 - \alpha$ is given by

$$CI = [C_0 - I, C_0 + I]$$

$$I = \sigma_t \times t\left(\frac{\alpha}{2}; n - 2\right) \quad (5)$$

In the current study, we employ 0.01 for the value of α .

2.4 Costs for the translation paired comparison method

The translation paired comparison method is an effective evaluation method because it can clearly express a system's performance as a TOEIC (Test of English for International Communication) score. However, this method has excessive evaluation costs.

Roughly speaking, one of these costs is the need to collect translations made by examinees of various TOEIC scores. As shown in Equations (4) and (5), n , the number of examinees, affects the confidence interval of the system's TOEIC score. Therefore, a reduction in this number makes it difficult to obtain a reliable evaluation result.

The other cost is for the evaluation. Compared to a conventional evaluation method, such as a simple rank evaluation method, the translation paired comparison method uses a larger amount of labor because the evaluator must work on n evaluation sheets. Each sheet consists of 330 pairs of translation results to be evaluated. Even for an accomplished evaluator, it takes more than two weeks to finish the work, following the method explained in section 2.2.

3 Proposed method

As explained in the previous section, the translation paired comparison method has an excessive evaluation cost. Nevertheless, it is an effective evaluation method for measuring the capability of a speech translation system. Therefore, cost reduction for this evaluation method is an important subject for study.

The proposed method reduces the evaluation cost by removing insensitive test utterances from the test set. In this section, we explain the optimization procedure of the proposed method.

3.1 Optimization basis

In the proposed method, the basis of test set optimization is the minimization of σ . As shown in Equations (4) and (5), this value has an influence on the confidence interval of the system's TOEIC score. Therefore, minimizing σ brings about a reliable evaluation result.

We introduce $\sigma_{iteration}$, which is calculated in each iteration step. $\sigma_{iteration}$ is also calculated by using Equations (2) and (3). The difference between $\sigma_{iteration}$ and σ is the test set to be used for calculation. $\sigma_{iteration}$ is calculated using residual test utterances in each iteration step. However, the values of β_1 and β_2 are fixed, i.e., for the calculation of $\sigma_{iteration}$, these β_1 and β_2 are calculated using the original test set consisting of 330 test utterances.

Optimization is conducted iteratively by picking up the test utterance that causes maximum $\sigma_{iteration}$ in each iteration step. The details of this procedure is explained in the next subsection.

3.2 Methodology of the proposed method

Figure 4 shows a diagram of the proposed method. In the first step, the number of iterations is set. This number is an actual number of removed test utterances. During the iterations, test utterances are removed one-by-one. To decide which test utterance to remove in each iteration, $\sigma_{iteration}$ is calculated for the condition of removing each test utterance. This calculation is done for all candidates, i.e., all constituents of residual test utterances.

At the end of each iteration step, the test utterance to be removed is decided. The removed

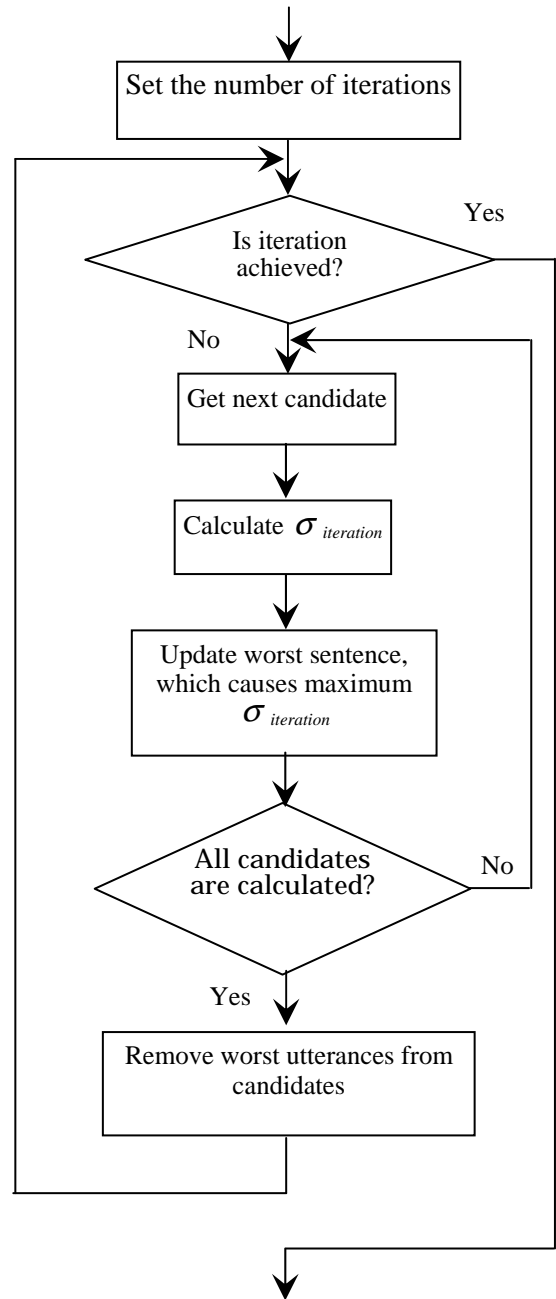


Figure 4: Procedure of proposed method

test utterance is the one that maximizes $\sigma_{iteration}$. We regard the utterance as maximizing $\sigma_{iteration}$ if removing it from the test set gives minimum $\sigma_{iteration}$.

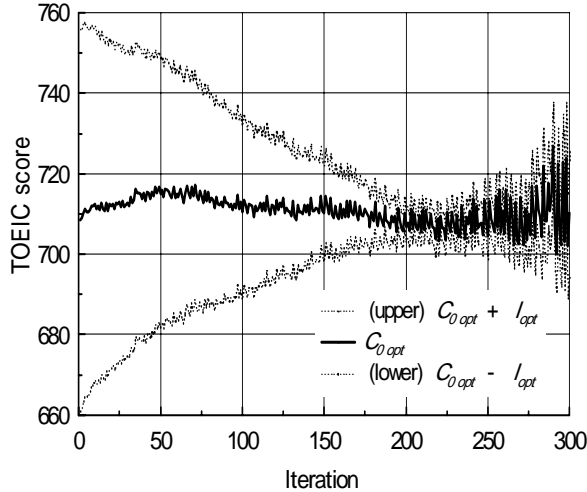


Figure 5: Relationship between iteration and system's TOEIC score

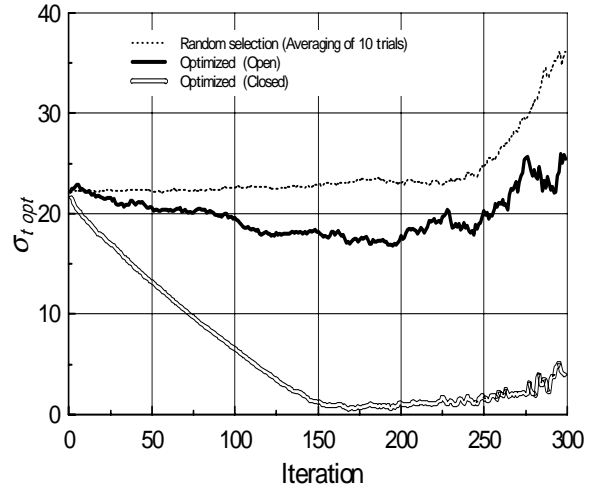


Figure 6: Relationship between iteration and $\sigma_{t\ opt}$

4 Experimental results

In this section, we show experimental results of the proposed method. Here, we introduce the suffix “*opt*” to distinguish a variable calculated with the optimized test set from a variable calculated with the original test set. All of the above variables are calculated with the original test set. By joining the suffix “*opt*” to these variables, we refer to variables calculated with the optimized test set, e.g., σ_{opt} ³, $\sigma_{t\ opt}$, I_{opt} , $C_{0\ opt}$, CI_{opt} , and so on.

4.1 Closed experiment

This subsection discusses an experimental result obtained for the same test set and examinees described in Section 2. Namely, the target test set for optimization consists of 330 utterances and the number of examinees is 30.

Figure 5 shows the relationship between iteration and the system's TOEIC score ($C_{0\ opt}$). In this figure, the horizontal axis represents the iteration number and the vertical axis the TOEIC score. The solid line represents $C_{0\ opt}$, which is the system's TOEIC score using the optimized test in each iteration. The dotted line above the solid line represents the value of $C_{0\ opt} + I_{opt}$, and the dotted line below the solid line $C_{0\ opt} - I_{opt}$.

As shown in the figure, from iteration 1 to iteration 250, the value of $C_{0\ opt}$ is stable and does not deviate from C_0 , which is 708. Furthermore, until around iteration 200, the value of I_{opt} decreases concurrently with the iteration.

This result suggests that the proposed may provide low-cost evaluation with high reliability.

4.2 Experiment opened for examinees

In the result shown in the previous subsection, the optimization and evaluation were conducted on the same examinees, i.e., the evaluation is closed for examinees. In this subsection, we look into the robustness of the proposed method against different examinees. We divided the group, consisting of 30 examinees, into two groups: a group of odd-numbered examinees and a group of even-numbered examinees. Individuals were sorted by TOEIC score from lowest to highest.

One of the groups is used to optimize the test set. The other group is used for the translation paired comparison method. We use the term “optimization group” to refer to the first group and “evaluation group” to refer to the second group.

Figure 6 shows the relationship between iteration and $\sigma_{t\ opt}$. In this figure, the horizontal axis represents the iteration and the vertical axis shows $\sigma_{t\ opt}$. Three kinds of experimental results are shown in this figure. In each of three experiments, the translation paired comparison is conducted by the evaluation group. The differences

³ σ_{opt} is different from $\sigma_{iteration}$. σ_{opt} is calculated based on $\beta_{1\ opt}$ and $\beta_{2\ opt}$ (not β_1 and β_2) for the optimized test set.

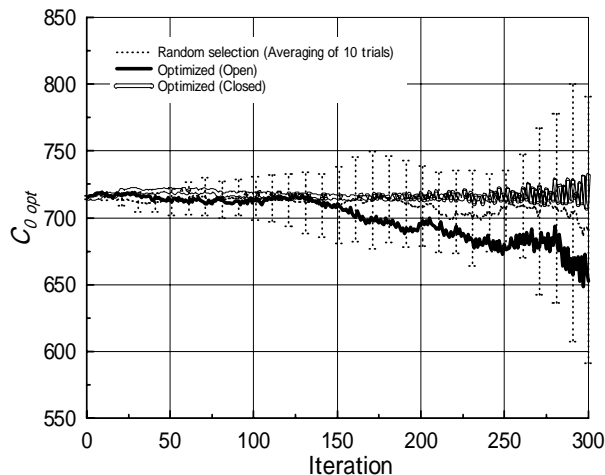


Figure 7: Relationship between iteration and $C_{0\ opt}$

among the three experiments are in the group to be used for optimization of the test set or the method used to reduce it. The double line represents the closed result using the test set, optimized on the evaluation group. The solid line represents the open result using the test set, optimized on the optimization group. The broken line represents the result using the test set, which is reduced by randomly removing test utterances one-by-one. The actual plotted broken line is averaged over 10 random trials.

As shown in Figure 6, in the random selection result, $\sigma_{t\ opt}$ is on the rise. On the other hand, the open result is on the decline.

Figure 7 shows the relationship between iteration and the system's TOEIC score. In this figure, the horizontal axis represents the iteration and the vertical axis the TOEIC score. The denotation of each line is the same as that in Figure 6. The error bar from the broken line represents σ_{random} , which is the standard deviation of the system's TOEIC score over 10 random trials.

In Figure 7, considering σ_{random} , $C_{0\ opt}$ of the open evaluation is more approximate to C_0 than that of random selection, whereas $C_{0\ opt}$ of the closed evaluation is much more approximate to C_0 .

4.3 Experiment on ATR-MATRIX

To be of actual use, the test set optimized for some system must be applicable for evaluation of other systems. In this subsection, we show the results of an experiment aimed at verifying this requirement is met. In this experiment, we apply the test set,

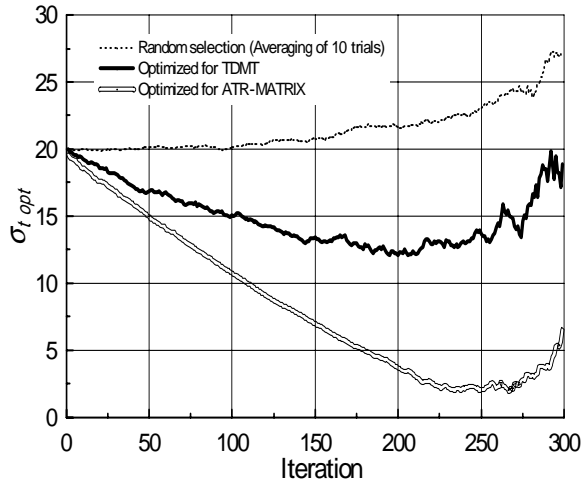


Figure 8: Relationship between iteration and $\sigma_{t\ opt}$

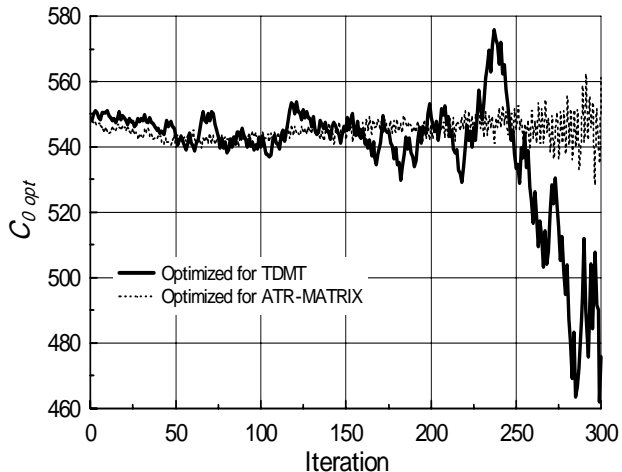


Figure 9: Relationship between iteration and $C_{0\ opt}$

which is optimized for TDMT, to evaluate ATR-MATRIX. The experimental conditions are the same as in Section 4.1, except for the evaluation target. The results are shown in Figure 8 and Figure 9.

Figure 8 shows the relationship between iteration and $\sigma_{t\ opt}$. In this figure, the horizontal axis represents the iteration and the vertical axis shows $\sigma_{t\ opt}$. The double line represents the result using the test set, optimized for ATR-MATRIX. The solid line represents the result using the test set, optimized for TDMT. The broken line represents the result using the test set, which is reduced by randomly removing test utterances one-

by-one. The actual plotted broken line is averaged over 10 random trials.

Figure 9 shows the relationship between iteration and the system's TOEIC score. In this figure, the horizontal axis represents the iteration, and the vertical axis TOEIC score. The broken line and the solid line are plotted using the same denotation as that in Figure 8.

In Figure 8, the solid line always lies on a lower position than the broken line. In Figure 9, from iteration 1 to around iteration 200, the broken line does not deviate from the actual system's TOEIC score, which is 548.

Considering these results, the test set optimized for TDMT is shown to be applicable for evaluating ATR-MATRIX.

5 Discussion

In this section, we discuss the experimental results shown in Section 4.

Looking at the broken lines in Figure 6 and Figure 8, test set reduction using random selection always causes an increase of $\sigma_{t\ opt}$ i.e., an increase in the scale of confidence interval. Therefore, this method causes the reliability of the evaluation result to deteriorate. Meanwhile, in the case of using the proposed method, looking at the solid lines on these figures, $\sigma_{t\ opt}$ is on the decline until around iteration 200. This means that we can achieve a more reliable evaluation result with a lower evaluation cost than when using the original test set. Here, looking at the solid lines in Figure 7 and Figure 9, the $C_{o\ opt}$ system's TOEIC score is nearly stable until iteration 200, and it does not deviate from C_o . As mentioned before, C_o for Figure 7 is 708 and C_o for Figure 9 is 548.

Considering these results, the proposed method can reduce the 330-utterance test set to a 130-utterance test set while reducing the scale of confidence interval. In other words, the proposed method both reduces evaluation costs by 60% and improves reliability of the evaluation result.

Looking at Equations (4) and (5), the scale of confidence interval is also influenced by n . When we allow the scale of confidence interval obtained from the original test set, we can use the proposed method's reduction effect of σ_t to compensate the σ_t 's increase by reducing n . In this case, the actual achievable cost reduction will be more than 60%.

6 Conclusions

We proposed a test set selection method for evaluating a speech translation system. This method optimizes and drastically reduces the test set required by the translation paired comparison method.

Translation paired comparison is an effective method for measuring a system's performance as a TOEIC score. However, this method has excessive evaluation costs. Therefore, cost reduction for this evaluation method is an important subject for study. We applied the proposed method in an evaluation of ATR-MATRIX. Experimental results showed the effectiveness of the proposed method. This method reduced evaluation costs by more than 60% and also improved the reliability of the evaluation result.

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References

- Morimoto, T., Uratani, N., Takezawa, T., Furuse, O., Sobashima, Y., Iida, H., Nakamura, A., Sagisaka, Y., Higuchi, N. and Yamazaki, Y. 1994. A speech and language database for speech translation research. In *Proceedings of ICSLP '94*, pages 1791-1794.
- Sugaya, F., Takezawa, T., Yokoo, A., Sagisaka, Y. and Yamamoto, S. 2000. Evaluation of the ATR-MATRIX Speech Translation System with a Pair Comparison Method between the System and Humans. In *Proceedings of ICSLP 2000*, pages 1105-1108.
- Sumita, E., Yamada, S., Yamamoto K., Paul, M., Kashioka, H., Ishikawa, K. and Shirai, S. 1999. Solutions to Problems Inherent in Spoken-language Translation: The ATR-MATRIX Approach. In *Proceedings of MT Summit '99*, pages 229-235.
- Takezawa, T. 1999. Building a bilingual travel conversation database for speech recognition

research. In *Proceedings of Oriental COCOSDA Workshop*, pages 17-20.

Takezawa, T., Morimoto, T., Sagisaka, Y., Campbell, N., Iida, H., Sugaya, F., Yokoo, A. and Yamamoto, S. 1998. A Japanese-to-English speech translation system: ATR-MATRIX. In *Proceedings of ICSLP 1998*, pages 2779-2782.