# **Bleu Score Study**

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# ABSTRACT

Machine translation (MT) is a fast growing sub-field of computational linguistic. Until now, the most popular automatic metrics to measure the quality of MT is Bleu score. Lately, MT along with its Bleu metric has been applied to many Software Engineering(SE) tasks. In this paper, we studied Bleu score to validate its suitability for software engineering tasks. We showed that Bleu score does not reflect translation quality due to its weak relation with semantic meaning of the translated source codes. Specifically, an increase in Bleu score does not guarantee an improved in translation quality, and a good translation may have fluctuated Bleu score. <sup>1</sup>

# **KEYWORDS**

ACM proceedings, LATEX, text tagging

#### **ACM Reference Format:**

#### 1 INTRODUCTION

Machine Translation (MT) is the use of computer program to translate text or speech from one language to another. The most popular automatic metrics to evaluate quality of MT is Bleu score. Traditionally, MT is only applied to natural language, but now it is also used for technical and programming language. One notable usage of MT for SE tasks is code migration. Even with that adapation, SE community still relies on Blue score. This leads to a question as whether Blue score is suitable for SE tasks or not.

Bleu score evaluate the quality of MT by calculating the modified n-grams precision and also taking into account the length difference penalty. Bleu was proved to be correlated with human judgments in natural language MT systems [?]. However, Callison at el argued that we should not over-rely on Bleu score as an improvement in Bleu score is not sufficient nor necessary to show an in improvement in translation

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© 2018 Association for Computing Machinery. ACM ISBN 978-x-xxxx-xxxx-x/YY/MM...\$15.00 https://doi.org/10.1145/nnnnnn.nnnnnnn quality [1]. To validate the use of Bleu on SE tasks, we set up an experiment to manually judge the result of multiple MT systems and compare its to the Bleu score. Our result showed that Bleu score has weak correlation to human judgments across

# 2 BACKGROUND

Background information and metrics formulation

#### 2.1 Bleu Score

Bleu (bilingual evaluation understudy) uses the modified form of n-grams precision and length difference penalty to evaluate the quality of text generated by MT compared to referenced one.

- 2.2 Lexical Score
- 2.3 Syntax Score
- 2.4 Semantic Score

#### 3 HYPOTHESIS AND EXAMPLES

Our hypothesis is that bleu score does not measure well the closeness in term of semantics between the reference and translated source code.

- 3.1 Bleu score is not sufficient
- 3.2 Bleu score is not necessary

#### 4 EMPERICAL STUDY

We have two RQs:

RQ1: Does bleu score reflect semantic meaning of translated source code?

RQ2: If Bleu score does not reflect semantic meaning, to what extend of source code it represents?

#### 4.1 RQ1

We study the correlation between bleu score and semantic score. Our results show that Bleu score and Semantic score have weak correlation.

## 4.2 RQ2

If the answer to RQ1 is 'no', we would like to investigate to what extend of source code does bleu score represents.

# 5 APPROACH AND EVALUATION

Since Bleu score is not suitable for SE task (code migration), we propose a new metric RUBY to evaluate quality of machine translation.

<sup>&</sup>lt;sup>1</sup>More abstract

#### 6 RELATED WORKS

## 7 CONCLUSIONS

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This paragraph will end the body of this sample document. Remember that you might still have Acknowledgments or Appendices; brief samples of these follow. There is still the Bibliography to deal with; and we will make a disclaimer about that here: with the exception of the reference to the LATEX book, the citations in this paper are to articles which have nothing to do with the present subject and are used as examples only.

#### REFERENCES

Chris Callison-Burch, Miles Osborne, and Philipp Koehn. 2006.
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