

Bilingual Evaluation Understudy (BLEU)

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BLEU idea

unigram precision

Modified- n-gram precision

Combining n-gram precisions

Demo

Other Metrics

DIFFICULTIES WITH HUMAN EVALUATION OF MT

- ▶ Human evaluations are extensive but expensive
- ▶ A need for quick, reusable, inexpensive method that correlates highly with human evaluation
- ▶ Many aspects of translation, including adequacy and fluency should be considered during the automatic evaluation
- ▶ Automatic evaluation is a boon to developers of MT
- ▶ Two important aspects required for automatic evaluation
 1. A good metric
 2. A good/gold standards as references

THE IDEA

- ▶ Many translations possible for a given sentence
- ▶ A good translator identifies a good candidate using adequacy and fluency

The main idea is to use a weighted average of variable length phrase matches against the reference translations[1]

Candidate 1: **It is a guide to action which ensures that the military always obeys the commands of the party**

Candidate 2: **It is to insure the troops forever hearing the activity guidebook that party direct**

Reference: **It is a guide to action that ensures that the military will for ever heed Party commands**

If many words and phrases are shared between the candidate and the reference translations, then it a good choice

Can n-grams help in matching the words and phrases?

UNIGRAM PRECISION

C1: It is a guide to action which ensures that the military always **obeys** the commands of the party

R1: It is a guide to action that ensures that the military will forever heed Party commands

R2: It is the guiding principle which guarantees the military forces always being under the command of the Party.

R3: It is the practical guide for the army to heed the directions of the party.

$$\text{Unigram precision} = \frac{17}{18}$$

C2: It is to **insure** the **troops** forever **hearing** the **activity guidebook** that party **direct**

R1: It is a guide to action that ensures that the military will forever heed Party commands

R2: It is the guiding principle which guarantees the military forces always being under the command of the Party.

R3: It is the practical guide for the army always to heed the directions of the party.

$$\text{Unigram precision} = \frac{8}{14}$$

MODIFIED- N-GRAM PRECISION

Compare the number of n-grams in the candidate and in the reference translation

Penalize models that produces many words of the same type

- ▶ Count the number of times a word occurs in any single reference translation
- ▶ $\text{Count}_{\text{clip}} = \min(\text{Candidate Count}, \text{Maximum Reference Count})$

Refer the previous slide for the examples

Modified unigram precision for C1 = $\frac{17}{18}$

• Modified unigram precision for C2 = $\frac{8}{14}$

C3: the the the the the the the

R4: the cat is on the mat

Unigram precision = $\frac{7}{7}$

Modified unigram precision = $\frac{2}{7}$

Modified bigram precision = 0

Modified Unigram precision defines the adequacy of the translation, while modified bigram precision matches the fluency of the translation

MODIFIED BIGRAM PRECISION

(It,is),(is,a),(a,guide),
(guide,to),(to,action),
(action,which),(which,ensures),
(ensures,that),(that,the),
(the,military),(military,always),
(always,obeys),(obeys,the),
(the,commands),(commands,of),
(of,the),(the,party)

Modified bigram precision for C1 = $\frac{10}{17}$

(It,is),(is,a),(a,guide),(guide,to),
(to,action),(action,that),(that,ensures),
(ensures,that),(that,the),(the,military),
(military,will),(will,forever),(forever,heed),
(heed,Party),(Party,commands)

(It,is),(is,the),(the,guiding),
(guiding,principle),(principle,which),
(which,guarantees),(guarantees,the),
(the,military),(military,forces),(forces,always),
(always,being),(being,under),(under,the),
(the,command),(command,of),
(of,the),(the,Party)

(It,is),(is,the),(the,practical),(practical,guide),
(guide,for),(for,the),(the,army),
(army,always),(always,to),(to,heed),
(heed,the),(the,directions),
(directions,of),(of,the),(the,party)

MODIFIED BIGRAM PRECISION - CANDIDATE 2

(It,is),(is,to),(to,insure),
(insure,the),(the,troops),
(troops,forever),(forever,hearing),
(hearing,the),(the,activity),
(activity,guidebook),
(guidebook,that),(that,party),
(party,direct)

Modified bigram precision for C2 = $\frac{1}{13}$

(It,is),(is,a),(a,guide),(guide,to),
(to,action),(action,that),(that,ensures),
(ensures,that),(that,the),(the,military),
(military,will),(will,forever),(forever,heed),
(heed,Party),(Party,commands)

(It,is),(is,the),(the,guiding),
(guiding,principle),(principle,which),
(which,guarantees),(guarantees,the),
(the,military),(military,forces),(forces,always),
(always,being),(being,under),(under,the),
(the,command),(command,of),
(of,the),(the,Party)

(It,is),(is,the),(the,practical),(practical,guide),
(guide,for),(for,the),(the,army),
(army,always),(always,to),(to,heed),
(heed,the),(the,directions),
(directions,of),(of,the),(the,party)

COMBINING N-GRAM PRECISIONS

- ▶ Modified n-gram precisions decay exponentially as n increases[1]
- ▶ BLEU uses a average log with a uniform weights to tackle the decay problem to get a score equivalent to the geometric mean of modified n-gram precisions
- ▶ $c < r$ inflates the precision
- ▶ A brevity penalty (BP) is introduced when $c \leq r$

$$BP = \begin{cases} 1, & \text{if } c > r \\ \exp(1 - \frac{r}{c}), & \text{if } c \leq r \end{cases}$$

where r is the effective length of the reference corpus and c is the length of the candidate sentence

BLEU score is obtained by

$$\text{BLEU} = \text{BP} \cdot \exp \sum_{n=1}^N w_n \log p_n \quad (1)$$

where N is the n -gram size (BLEU uses 4-gram by default), w_n is the weights associated with unigram, bigram, trigram and 4-grams, and p_n is the modified precision score of the test corpus. Here, $\sum_{n=1}^N w_n = 1$. One option for $w_n = \frac{1}{N}$

$$p_n = \frac{\sum_{c \in C} \sum_{n\text{grams} \in C} \text{Count}_{\text{clip}}(n\text{grams})}{\sum_{c \in C} \sum_{n\text{grams} \in C} \text{Count}(n\text{grams})} \quad (2)$$

BLEU Demo

BLEU is designed as a corpus measure

- ▶ Machine translation
- ▶ Image labeling
- ▶ Text summarization
- ▶ Speech recognition

OTHER METRICS

- ▶ NIST - National Institute of Standards and Technology - based on BLEU
- ▶ METEOR - Metric for Evaluation of Translation with Explicit ORdering
 - Uses stemming and synonymy matching
- ▶ WER - Word Error Rate
 - ▶ Uses edit distance (Levenshtein distance)
 - ▶ Finds minimum number of edit operations such as insertion, deletions or substitutions, needed to change the candidate sentence into the reference sentence
- ▶ GLEU - Google BLEU
 - ▶ Correlates well with BLEU, and works with sentence level translation

REFERENCES

- [1] Kishore Papineni et al. “Bleu: a Method for Automatic Evaluation of Machine Translation”. In: *Proceedings of 40th Annual Meeting of the Association for Computational Linguistics*. Philadelphia, Pennsylvania, USA: Association for Computational Linguistics, July 2002, pp. 311–318. DOI: [10.3115/1073083.1073135](https://doi.org/10.3115/1073083.1073135). URL: <https://www.aclweb.org/anthology/P02-1040>.