

BLEU Score Evaluation Metric

♦ What is BLEU Score?

BLEU (Bilingual Evaluation Understudy) is a quantitative metric to evaluate how close a machine-generated sentence is to a human reference translation. It's one of the most widely used metrics in machine translation and text generation tasks.

♦ Two Key Components of BLEU

1. N-gram Precision

Measures how many n-grams (sequences of n words) in the generated output match with the reference translation. It is computed for multiple n-gram levels (unigrams, bigrams, trigrams, etc.).

2. Brevity Penalty (BP)

Prevents models from cheating by generating shorter sentences to get high precision.

♦ BLEU Score Formula

$$\text{BLEU} = \text{BP} \times \exp\left(\sum w_n \times \log(P_n)\right)$$

Where:

- P_n = modified precision for n-grams
- w_n = weight for n-gram (e.g., 0.25 for uniform 1–4 grams)
- $\text{BP} = 1$ if $c > r$, else $\exp(1 - r/c)$
where c = candidate length, r = reference length

♦ Example

Reference Sentence: "a cat is sitting on the mat"

Candidate Sentence: "cat is sitting on mat"

♦ Step-by-step Calculation

1. Unigram Precision

Reference unigrams: ["a", "cat", "is", "sitting", "on", "the", "mat"]

Candidate unigrams: ["cat", "is", "sitting", "on", "mat"]

Matched: $5 / 5 = 1.0$

Precision $P_1 = 1.0$

2. Bigram Precision

Reference bigrams: ["a cat", "cat is", "is sitting", "sitting on", "on the", "the mat"]

Candidate bigrams: ["cat is", "is sitting", "sitting on", "on mat"]

Matched: $3 / 4 = 0.75$

Precision $P_2 = 0.75$

3. Trigram Precision

Reference trigrams: ["a cat is", "cat is sitting", "is sitting on", "sitting on the", "on the mat"]

Candidate trigrams: ["cat is sitting", "is sitting on", "sitting on mat"]

Matched: $2 / 3 \approx 0.6667$

Precision $P_3 \approx 0.6667$

4. Brevity Penalty

$c = 5$ (candidate length), $r = 7$ (reference length)

$BP = \exp(1 - r/c) = \exp(1 - 7/5) = \exp(-0.4) \approx 0.6703$

5. Final BLEU Score

Using $n = 3$ (up to trigrams), uniform weights $w_n = 1/3$

$BLEU = 0.6703 \times \exp((1/3) \times (\log(1.0) + \log(0.75) + \log(0.6667)))$

$= 0.6703 \times \exp(-0.2311) = 0.6703 \times 0.7938 \approx 0.532$

♦ Python Libraries to Compute BLEU Score

- nltk.translate.bleu_score
- sacrebleu
- evaluate (from HuggingFace)

♦ Applications of BLEU

- Machine Translation
- Text Summarization
- Dialogue Generation
- Image Captioning

Summary

BLEU helps benchmark your model's linguistic precision, ensuring the generated sentences are not only accurate but also contextually complete.