EVALUATION METRICS FOR SPEECH TRANSLATION

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Overview

- Conventional Metrics
 - Standard Evaluation Metrics

- 2 End-to-End Metrics
 - Limitations and Research Challenges

Standard Evaluation Metrics

- Human Judgement(Subjective)
- Automated Evaluation Metrics(Objective)

The Conventional Approach

- Automatic Speech Recognition[ASR]-Word Error Rate [WER]
- Machine Translation[MT]- Automated Metrics
- Text to Speech Synthesis[TTS]- Mean Opinion Score[MOS]

Word Error Rate

$$WER = (S+I+D)/N$$

- S substitutions
- I- insertions
- D- deletions
- N- number of said words

Word Error Rate

ORIGINAL

Speech to Speech Translation System

WER=3/5=0.6

ASR-Output

Speech to Text(S) —(D) System Model(I)

Limitations

- The Metric for speech to text is not dynamic
- Works good only on domain specific Language
- Poor on Accents and Homophones

AUTOMATED METRICS

- BiLingual Evaluation Understudy (BLEU)
- Metric for Evaluation of Translation with Explicit word Ordering (METEOR)
- Translation Edit Rate (TER)
- Ref:Condon et al. "Automated metrics for speech translation." In Proceedings of the 9th Workshop on Performance Metrics for Intelligent Systems, 2009

BLEU

- Developed by IBM researchers in 2001
- BLEU score is to compare n-grams of the candidate with the n-grams of the reference translation
- The more the matches, the better the candidate translation

BLEU-Example

- Kannada sentence: "Budakattu bhaashegaligaagi Dhwanibhaashaanuvaada ghataka"
- Reference Sentence: "SPEECH TO SPEECH TRANSLATION MODULE FOR TRIBAL LANGUAGES"
- Candidate Sentence: "SPEECH TO SPEECH TRANSLATION"
- 1 gram- 4/4
- 2 gram- 3/3
- 3 gram- 2/2
- 4 gram- 1/1
- BLEU= Geometric mean of n grams= $\sqrt[4]{(1*1*1*1)}$

BREVITY PENALTY

- BP= 1 if c > r and BP= exp(1 r/c) if c < r
- BLEU = BP x GM
- $\exp(1-8/4) = \exp(-1) = 0.37(BLEU score)$

BLUE-Limitations

- The matches are position independent
- Measures Precision But Not Recall
- BP is not penalising repetitions
- Stemming and Synonyms are considered as zero match

The METEOR Measure

- Metric for Evaluation of Translation with Explicit word Ordering
- Developed by Carnegie Mellon University in 2005
- Considers Both Precision and Recall

The METEOR Measure

- Precision (P) = m/wt
- Recall (R) = m/wr
- Fmean = 10PR/(R+PR)
- m: number of unigrams in the candidate found in reference
- wt: Number of unigrams in candidate translation
- wr: Number of unigrams in reference translation

CHUNK PENALTY

- $p = 0.5(C/Um)^3$
- C: Number of chunks in candidate
- Um:Unigrams in candidate

The METEOR-Example

- Reference : SPEECH TO SPEECH TRANSLATION MODULE FOR TRIBAL LANGUAGES
- Candidate : SPEECH TO SPEECH TRANSLATION
- $p = 0.5 (1/4)^3 = 0.0078$
- P = 1, R = 0.5, Fmean = 0.53
- M = Fmean (1- p)=0.53(0.99)=0.525

TRANSLATION EDIT RATE

- Developed by University Of Maryland in 2006
- TER = (Substitutions + Insertions + Deletions + Shifts)/Reference Words
- Suitable for both Machine and Human Evaluation

MEAN OPINION SCORE

- This is a Human Evaluation technique.
- Measured on the 5-point scale for Adequacy, Fluency, Naturalness

Limitations

- Can't rely on MOS for Speech Translation system
- Synthesized voice does not consider the correctness of translation

Translatotron(2021)

Table 5: Multilingual X→En S2ST performance on 4 high-resource languages from CoVoST 2, measured by BLEU on ASR transcribed text. The same checkpoints from each model were used for evaluating all language pairs. Note: BLEU scores are not directly comparable between S2ST and ST.

| Source language | fr | de | es | ca |
|-------------------------------|--------------|--------------|--------------|--------------|
| Translatotron 2 Translatotron | 27.0 18.9 | 18.8 10.8 | 27.7 18.8 | 22.5 13.9 |
| ST (Wang et al., 2021a) | 27.0 | 18.9 | 28.0 | 23.9 |
| Training target | 82.1 | 86.0 | 85.1 | 89.3 |

Jia, Ye, et al. "Translatotron 2: Robust direct speech-to-speech translation." arXiv preprint arXiv:2107.08661(2021)

Speech Transformer (2021)

Table 4. BLEU and METEOR scores of speech-to-speech translation

| | Syntactic similar | | | | Syntactic distant | | | |
|-----------------------------------|-------------------|--------|----------|--------|-------------------|--------|----------|--------|
| Model | En to Es | | Ja to Ko | | En to Ja | | Ja to En | |
| | BLEU | METEOR | BLEU | METEOR | BLEU | METEOR | BLEU | METEOR |
| Baseline: Cascade (RNN) | 38.9 | 47.7 | 38.7 | 49.1 | 32.5 | 44.2 | 32.0 | 43.2 |
| Baseline: Cascade (Transformer) | 41.3 | 52.1 | 41.0 | 51.1 | 34.1 | 45.2 | 35.0 | 45.3 |
| Google (RNN) [4] | 38.8 | 48.2 | 39.1 | 49.9 | 33.2 | 45.5 | 34.2 | 45.0 |
| Google (Transformer) ¹ | 43.1 | 58.8 | 42.5 | 58.3 | 36.9 | 52.6 | 38.3 | 48.4 |
| Transcoder (Transformer) | 44.0 | 59.3 | 42.9 | 58.8 | 40.6 | 56.6 | 41.0 | 55.8 |

¹In this experiment we constructed a Google system using a Transformer network.

Kano, Takatomo et al. "Transformer-Based Direct Speech-To-Speech Translation with Transcoder." 2021 IEEE Spoken Language Technology Workshop (SLT). IEEE, 2021.

Conclusion

- The automated evaluation metrics do not understand Morphological, Semantic and Syntactic factors
- The performance of evaluation metrics depends on human judgments
- No standard Benchmark
- Evaluation Metrics is an Open Challenge

Research Challenges

- Evaluating translation combined with speech synthesis
- Single Metric which can measure the performance of End to End Systems
- Metric for target languages without standardized orthographies
- Effect of dialectal variation/accent variation on Evaluation Metrics

References

- Condon et al. "Automated metrics for speech translation." In Proceedings of the 9th Workshop on Performance Metrics for Intelligent Systems, 2009
- Le, Ngoc Tien. "Advanced quality measures for speech translation."
 PhD diss., Université Grenoble Alpes, 2018.
- Banerjee, Satanjeev, and Alon Lavie. "METEOR: An automatic metric for MT evaluation with improved correlation with human judgments." Proceedings of the acl workshop on intrinsic and extrinsic evaluation measures for machine translation and/or summarization. 2005
- Jia, Ye, et al. "Translatotron 2: Robust direct speech-to-speech translation." arXiv preprint arXiv:2107.08661(2021).

References

- Kano, Takatomo, Sakriani Sakti, and Satoshi Nakamura.
 "Transformer-Based Direct Speech-To-Speech Translation with Transcoder." 2021 IEEE Spoken Language Technology Workshop (SLT). IEEE, 2021.
- Papineni, Kishore, et al. "Bleu: a method for automatic evaluation of machine translation." Proceedings of the 40th annual meeting of the Association for Computational Linguistics. 2002
- Snover, Matthew, et al. "A study of translation edit rate with targeted human annotation." Proceedings of the 7th Conference of the Association for Machine Translation in the Americas: Technical Papers. 2006.
- https://www.cs.cmu.edu/ alavie/papers/GALE-book-Ch5.pdf
- https://languagelog.ldc.upenn.edu/nll/?p=193

Thank You