

Evaluation protocol

Definition and examples

Herman Kamper

Passage and grid evaluations

can: ga u la e ka ha
ref: ga ga u e ka hi hi
hyp: ga u la i ka hi ho

- ASR evaluation: Word error rate (WER) (`wer_ref_hyp`)
- EGRA MAE: Mean absolute error (MAE) between EGRA_ACC and ASR_EGRA_ACC (call this `MAE_EGRA_ACC`)
 - Bias baseline
- Finer grained (explained below):
 - Mistake error rate (MER)
 - P, R, F1 for substitution, insertion, deletion and all mistakes

Complete example

can: ga u la e ka ha
ref: ga ga u e ka hi hi
hyp: ga u la i ka hi ho

Align reference to canonical (A):

can:		ga	u		la	e		ka		ha
		I	C	C	D	C	C	I	S	
ref:	ga	ga	u			e		ka	hi	hi

- $\text{wer_can_ref} = \frac{S+D+I}{N} = \frac{1+1+2}{6} = \frac{4}{6}$
- $\text{acc_can_ref (EGRA_ACC)} = \frac{C}{N} = \frac{4}{6}$

can: ga u la e ka ha
ref: ga ga u e ka hi hi
hyp: ga u la i ka hi ho

Align hypothesis to canonical (B):

can: ga u la e ka ha
 C C C S C I S
hyp: ga u la i ka hi ho

- $\text{acc_can_hyp (ASR_EGRA_ACC)} = \frac{C}{N} = \frac{4}{6}$

can: ga u la e ka ha
ref: ga ga u e ka hi hi
hyp: ga u la i ka hi ho

Align hypothesis to reference:

ref: ga ga u e ka hi hi
C D C I S C C S
hyp: ga u la i ka hi ho

- $$\text{wer_ref_hyp} = \frac{S+D+I}{N} = \frac{2+1+1}{7} = \frac{4}{7} = 0.5714$$

Compare the two predictions:

- $$\text{MAE_EGRA_ACC} = | \text{acc_can_ref} - \text{acc_can_hyp} | = | \frac{4}{6} - \frac{4}{6} | = 0$$

Align CSID sequences from A and B :

A: ref_to_can:	I	C	C	D	C		C	I	S
	D	C	C	D	C	I	C	C	C
B: hyp_to_can:		C	C		C	S	C	I	S

- $\text{mer} = \frac{2+1}{8} = 0.375$

- Substitutions:

- Precision = $\frac{1}{2}$
 - Recall = $\frac{1}{1}$

- Insertions:

- Precision = $\frac{1}{1}$
 - Recall = $\frac{1}{2}$

- Deletions:

- Precision = $\frac{0}{0}$
 - Recall = $\frac{0}{1}$

- All mistakes:

- Precision = $\frac{2}{3}$
 - Recall = $\frac{2}{4}$

Isolated letters, syllables and non-words

- ASR evaluation: Word error rate (WER) (`wer_ref_hyp`)
- EGRA: Accuracy in correct / incorrect prediction
- Finer-grained:
 - P, R, F1 for correct mistake prediction
 - A mistake has a label 1
 - Scores for a majority baseline (predict everything as either correct or incorrect, whichever occurs most)

Example:

can: i

ref: i o

hyp: -

Summarising metrics in tables

- Passage and grid reading:
 - ASR WER: `wer_ref_hyp`
 - EGRA:
 - Correlation coefficient r (see next slide) between predicted (hyp-ref) and actual number of correct (can-ref) words per utterance (segment)
 - Scatter plot in code
 - MAE between correct counts
 - Finer grained:
 - Mistake error rate (MER)
 - Substitution P, R, F1
 - Insertion P, R, F1
 - Deletion P, R, F1
 - All mistakes P, R, F1
- Isolated letters, syllables and non-words:
 - ASR WER: `wer_ref_hyp`
 - EGRA accuracy = $\frac{TP+TN}{N}$
 - P, R, F1 for correct mistake prediction (label 1 = mistake)
 - P, R, F1 for majority baseline

Resources

- [Link to notebook](#)
- Correlation code:

```
import numpy as np

r = np.corrcoef(human_wcpm, automated_wcpm)[0, 1]
print("Correlation r:", r)

from sklearn.metrics import r2_score

R2 = r2_score(human_wcpm, automated_wcpm)
print("R2:", R2)

R2_from_r = r**2
print(R2_from_r)
```