

# A reliability study on the ACLEW corpora

## Experimental set-up

For studying the reliability of human annotators, and get a sense of their level of agreement, we asked to a second person to annodate a 1-mn long chunk from the daylong recording of each child. The purpose being to compare metrics obtained by these two annotators. This reliability study has been performed on SOD, WAR, ROS, TSE, ROW and BER corpus. We have 10 children by sub-corpora, for a total of 60 chunks of 1 minute.

We mapped all the labels into :

- CHI : for the key-child, the one wearing the recording device
- OCH : for other children
- MAL : for male speakers
- FEM : for female speakers
- OVL : for overlap
- SIL : for silence

## Performances metrics

### Identification Error Rate as a overall performance measure

One might want compare the level agreement of the two annotators as a function of the identification error rate. As a reminder, the identification error rate is computed as follow :

$$\text{identification error rate} = \frac{\text{false alarm} + \text{miss} + \text{confusion}}{\text{total}}$$

where :

false alarm is the duration of non-speech incorrectly classified as speech

miss is the duration of speech incorrectly classified as non-speech

confusion is the duration of speaker confusion (agreements on the fact that there's speech, but disagreement on the talker identity).

speech is the duration of speaker confusion (agreements on the fact that there's speech, but disagreement on the talker identity).

The two annotators obtained an identification error rate of 55.57% shared amongst a false alarm rate of 19.85%, a miss rate of 18.97% and a confusion of 16.75%.

Here's the per corpora identification error rate :

| corpora | ider     |
|---------|----------|
| BER     | 41.05884 |
| ROW     | 44.29150 |
| SOD     | 49.11254 |
| WAR     | 54.26772 |
| ALL     | 55.57232 |
| ROS     | 71.43684 |
| TSE     | 72.14068 |

### Best cases

The three best cases, for which the agreement was the highest were :

|                             | ider% | total | correct | correct% | fa   | fa%  | miss | miss% | conf | conf% | cc |
|-----------------------------|-------|-------|---------|----------|------|------|------|-------|------|-------|----|
| WAR_9398_005100_005160.rttm | 11.93 | 15.01 | 14.22   | 94.74    | 1.00 | 6.66 | 0.79 | 5.26  | 0    | 0     | W  |
| WAR_3528_006660_006720.rttm | 11.95 | 40.26 | 36.12   | 89.72    | 0.67 | 1.66 | 4.14 | 10.28 | 0    | 0     | W  |
| WAR_4995_026700_026760.rttm | 12.96 | 49.70 | 44.71   | 89.96    | 1.45 | 2.92 | 4.99 | 10.04 | 0    | 0     | W  |

### Worst cases

The three worst cases, for which the disagreement was the highest were :

|                             | ider%  | total | correct | correct% | fa    | fa%    | miss | miss% | conf  | conf% | cc |
|-----------------------------|--------|-------|---------|----------|-------|--------|------|-------|-------|-------|----|
| WAR_1130_023040_023100.rttm | 921.74 | 4.83  | 1.89    | 39.13    | 41.58 | 860.87 | 0.19 | 3.93  | 2.75  | 56.94 | W  |
| TSE_0643_020364_020424.rttm | 521.55 | 6.96  | 6.03    | 86.64    | 35.37 | 508.19 | 0.90 | 12.93 | 0.03  | 0.43  | W  |
| ROS_1299_004320_004380.rttm | 153.48 | 33.88 | 6.72    | 19.83    | 24.84 | 73.32  | 0.25 | 0.74  | 26.91 | 79.43 | W  |

### Detection Error Rate as a per-class performance measure

One can have a look at the per-class detection error rate defined as :

$$\text{detection error rate} = \frac{\text{false alarm} + \text{miss}}{\text{total}}$$

|     | ALL      | BER       | ROW      | WAR      | TSE      | ROS      | SOD       |
|-----|----------|-----------|----------|----------|----------|----------|-----------|
| CHI | 21.06832 | 28.445883 | 20.98293 | 12.25221 | 28.18321 | 39.13366 | 13.80426  |
| FEM | 29.10666 | 21.441932 | 35.15565 | 22.72983 | 24.01249 | 50.71766 | 23.06536  |
| MAL | 36.56353 | 19.737533 | 21.77419 | 35.19952 | 60.27861 | 47.57974 | 100.00000 |
| OCH | 43.24036 | 66.388175 | 29.27194 | 30.66376 | 48.86864 | 33.25165 | 59.52865  |
| ELE | 49.63453 | 6.909967  | 17.61876 | 82.54005 | NaN      | 66.80583 | NaN       |
| OVL | 50.02050 | 40.797218 | 36.26136 | 71.31915 | 72.94763 | 16.28559 | 64.08200  |

With no surprise, there's a high disagreement for classes such as the OVL one for which it is harder to tell when it starts and it ends exactly. The highest agreement is obtained for the CHI class for which the two annotators obtained a detection error rate of 40.27%

### Best agreement for the CHI class

The three best cases, for which the agreement on the CHI class was the highest were :

|                             | deter% | total | fa | fa% | miss | miss% |
|-----------------------------|--------|-------|----|-----|------|-------|
| BER_6035_030360_030420.rttm | 0      | 0     | 0  | NA  | 0    | NA    |
| BER_7758_034320_034380.rttm | 0      | 0     | 0  | NA  | 0    | NA    |
| ROS_3510_004740_004800.rttm | 0      | 0     | 0  | NA  | 0    | NA    |

### Worst agreement for the CHI class

The three worst cases, for which the disagreement on the CHI class was the highest were :

|                             | deter%  | total | fa   | fa%     | miss | miss%  |
|-----------------------------|---------|-------|------|---------|------|--------|
| TSE_7220_030589_030649.rttm | 1650.00 | 0.12  | 1.86 | 1550.00 | 0.12 | 100.00 |
| ROW_2745_020220_020280.rttm | 338.71  | 0.31  | 1.05 | 338.71  | 0.00 | 0.00   |
| SOD_1499_024300_024360.rttm | 226.12  | 1.34  | 2.05 | 152.99  | 0.98 | 73.13  |

### Precision/Recall as a per-class performance measure

As illustrated by the two tables shown above, the detection error rate (like the identification error rate) can be tricky to interpret when little speech is contained in the chunk. Indeed, in that particular case, the denominator is close to 0 (or equal to 0 if there's no speech), hence pumping up the measure. One might be more familiar with metrics such as the precision and the recall defined as :

$$\text{precision} = \frac{tp}{tp + fp}$$

$$\text{recall} = \frac{tp}{tp + fn}$$

where :  $tp$  is the duration of true positive (e.g. speech classified as speech)

$fp$  is the duration of false positive (e.g. non-speech classified as speech)

$fn$  is the duration of false negative (e.g speech classified as non-speech)

| class | precision | recall |
|-------|-----------|--------|
| CHI   | 80.44     | 78.93  |
| FEM   | 72.10     | 70.89  |
| MAL   | 72.05     | 63.44  |
| OCH   | 64.50     | 56.76  |
| ELE   | 50.35     | 50.37  |
| OVL   | 35.40     | 49.98  |

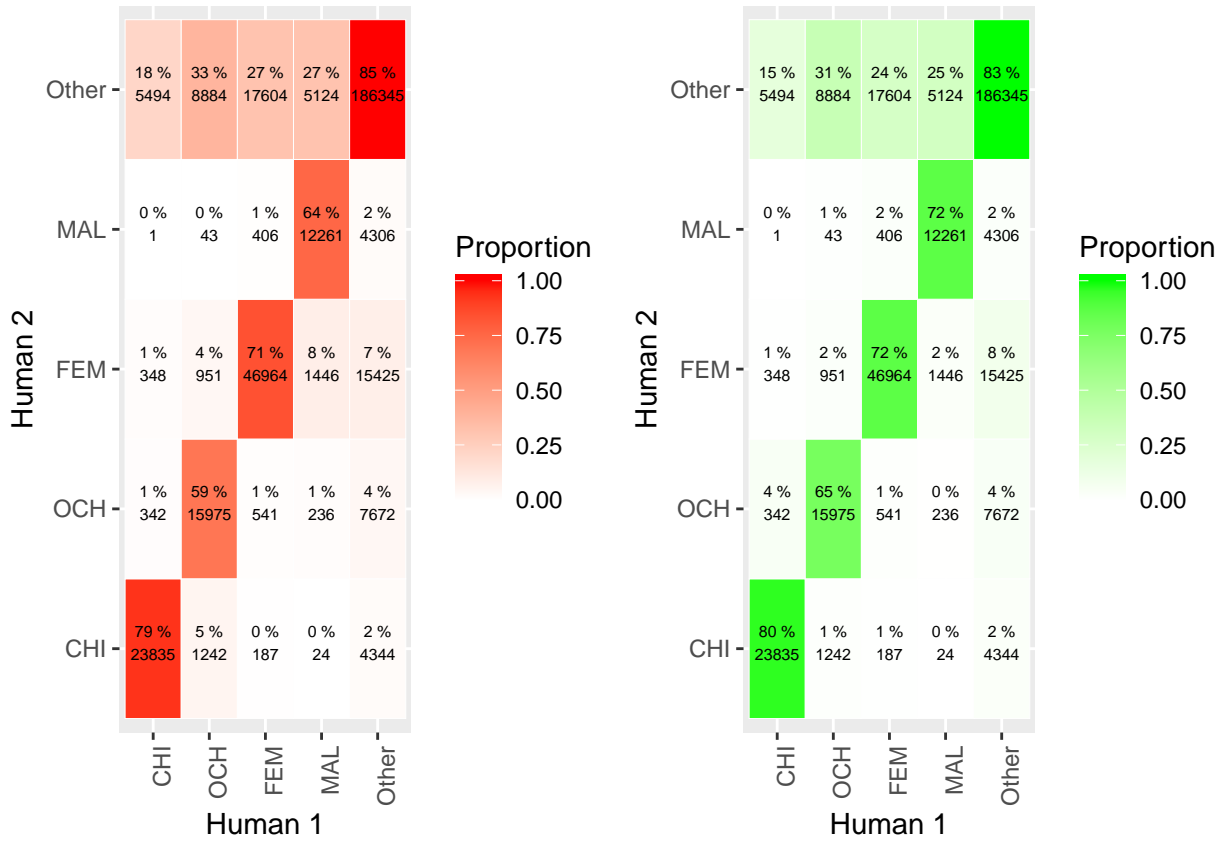
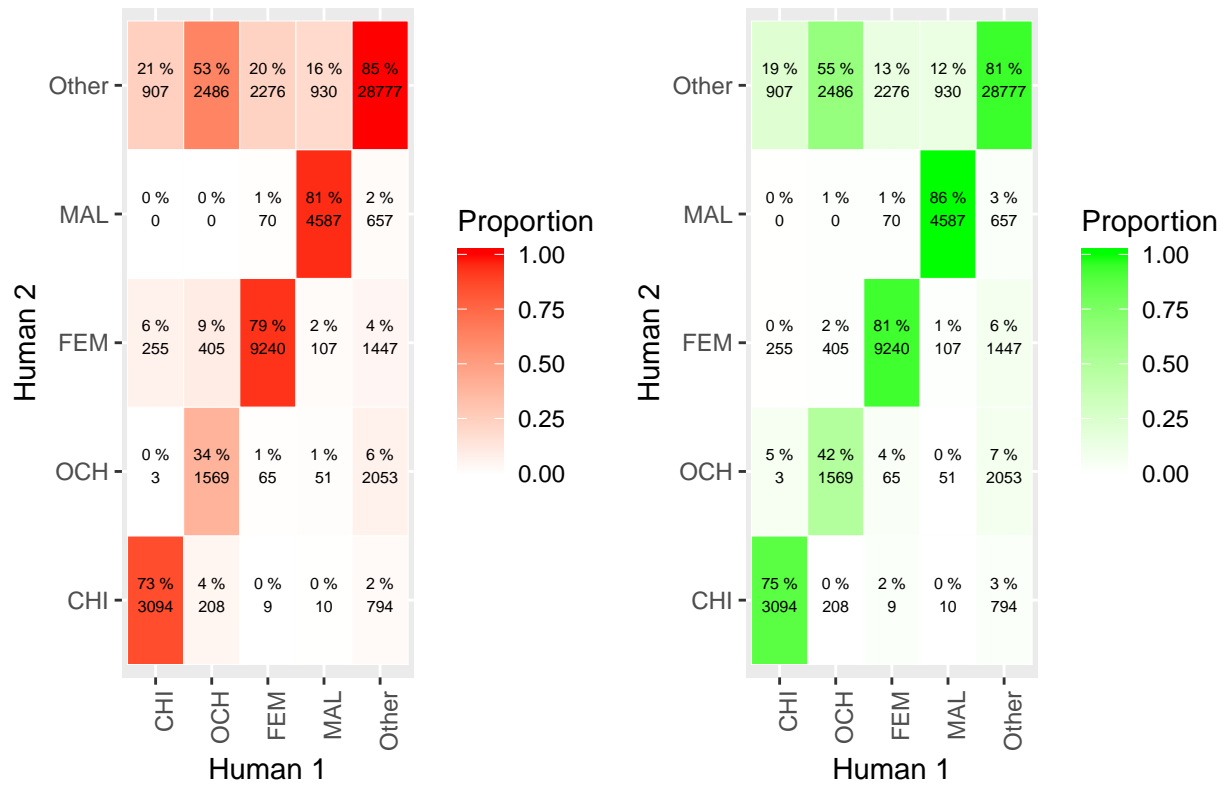
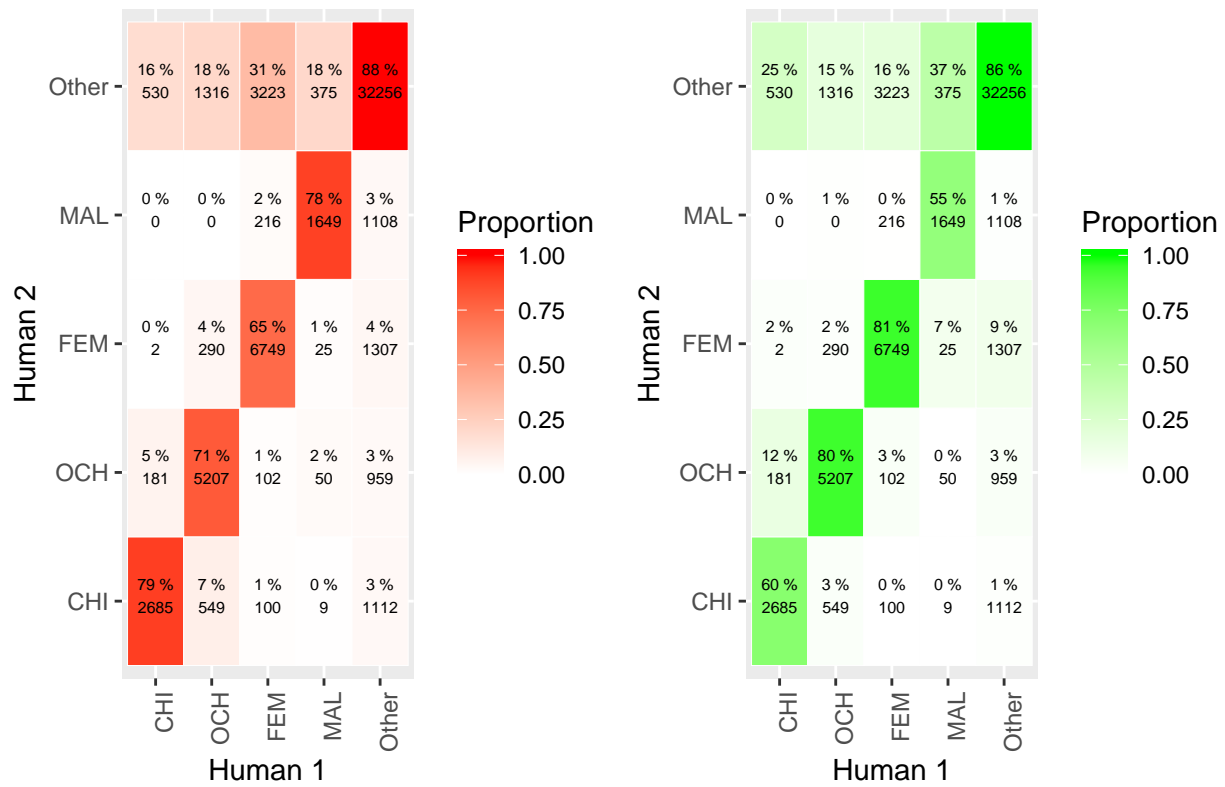


Figure 1: Precision (left) and recall (right) confusion matrices on all of the corpus

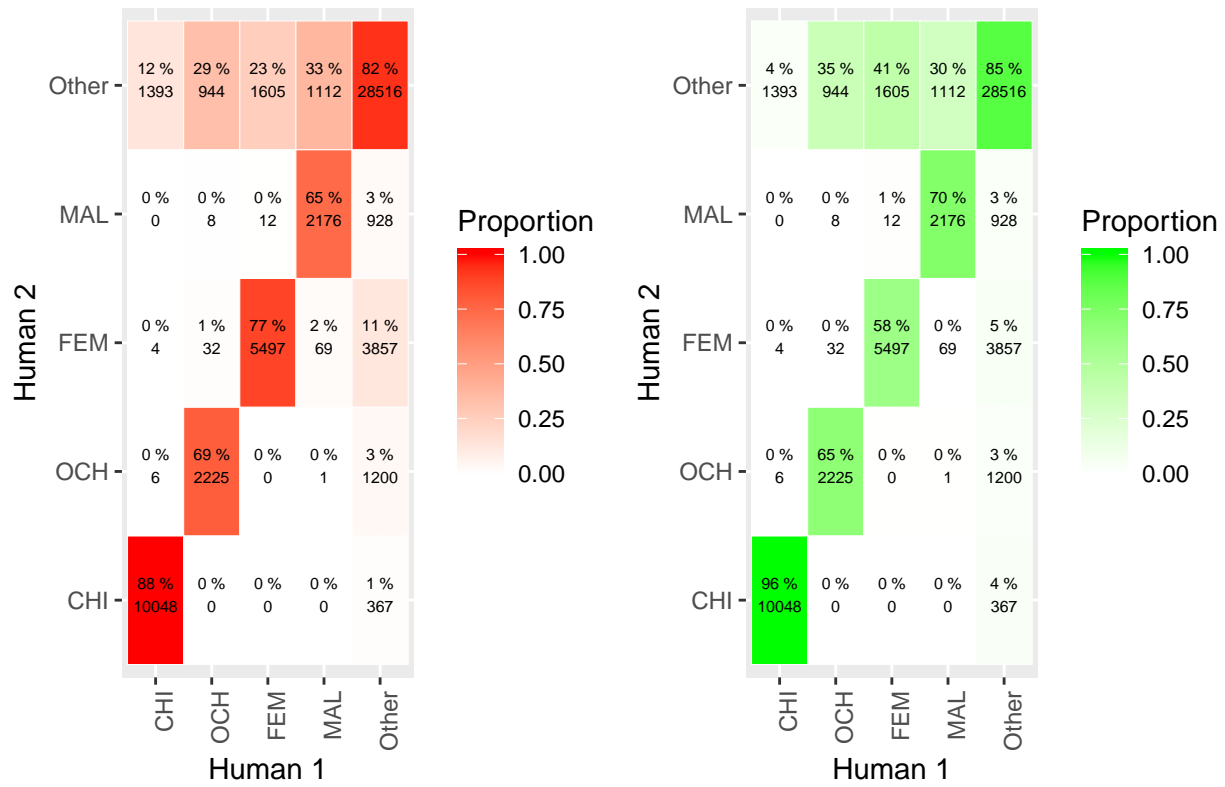
## Precision/Recall on BER



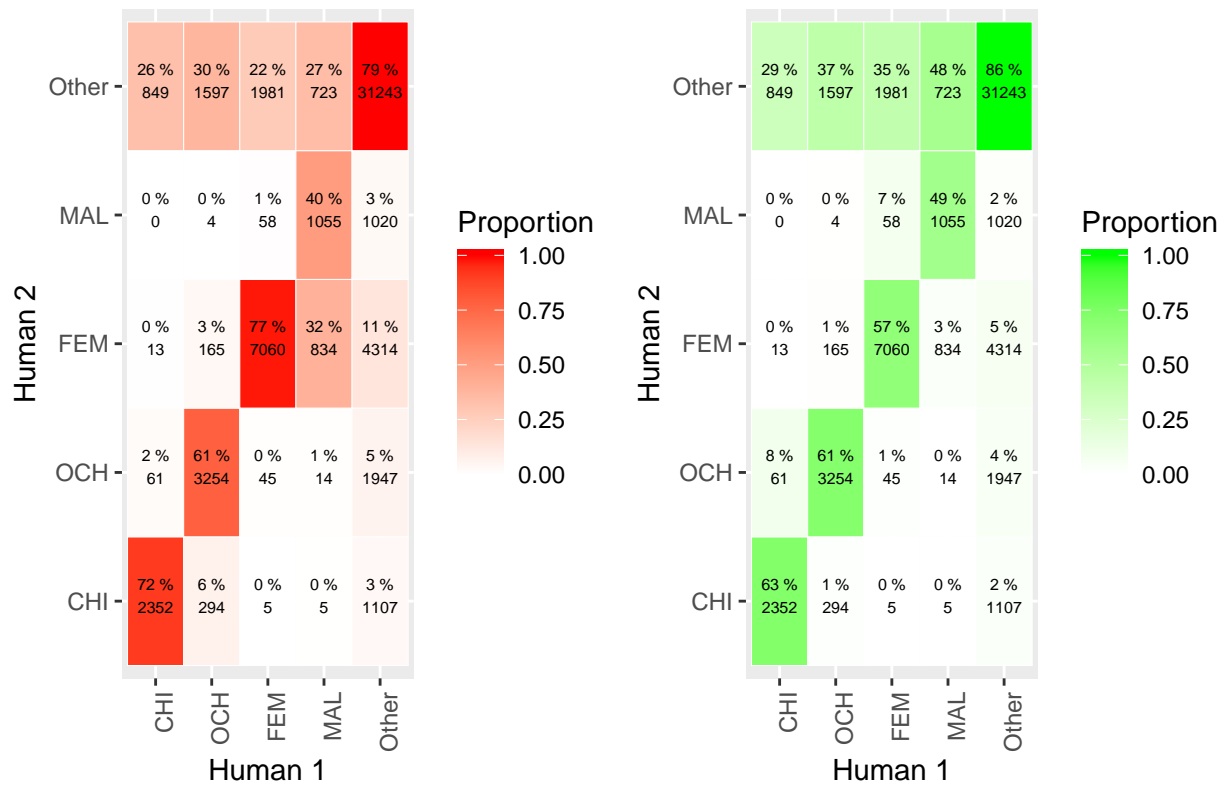
## Precision/Recall on ROW



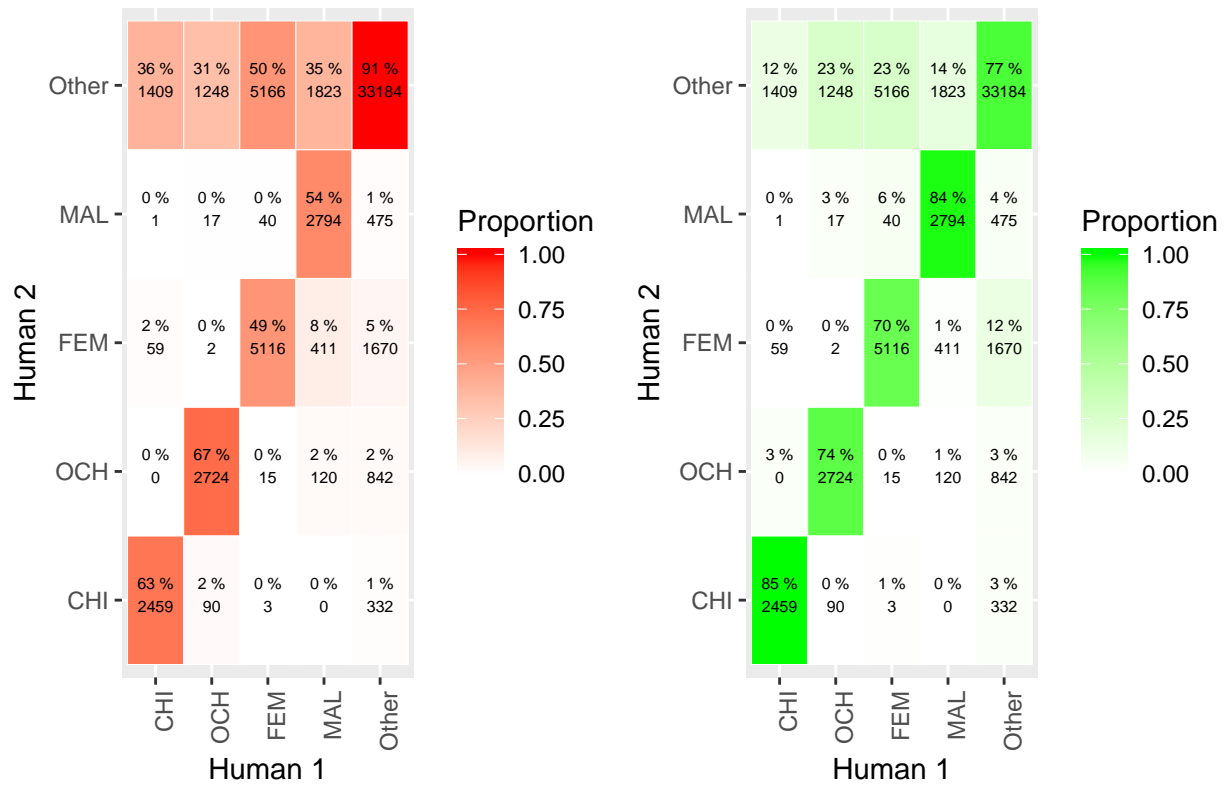
## Precision/Recall on WAR



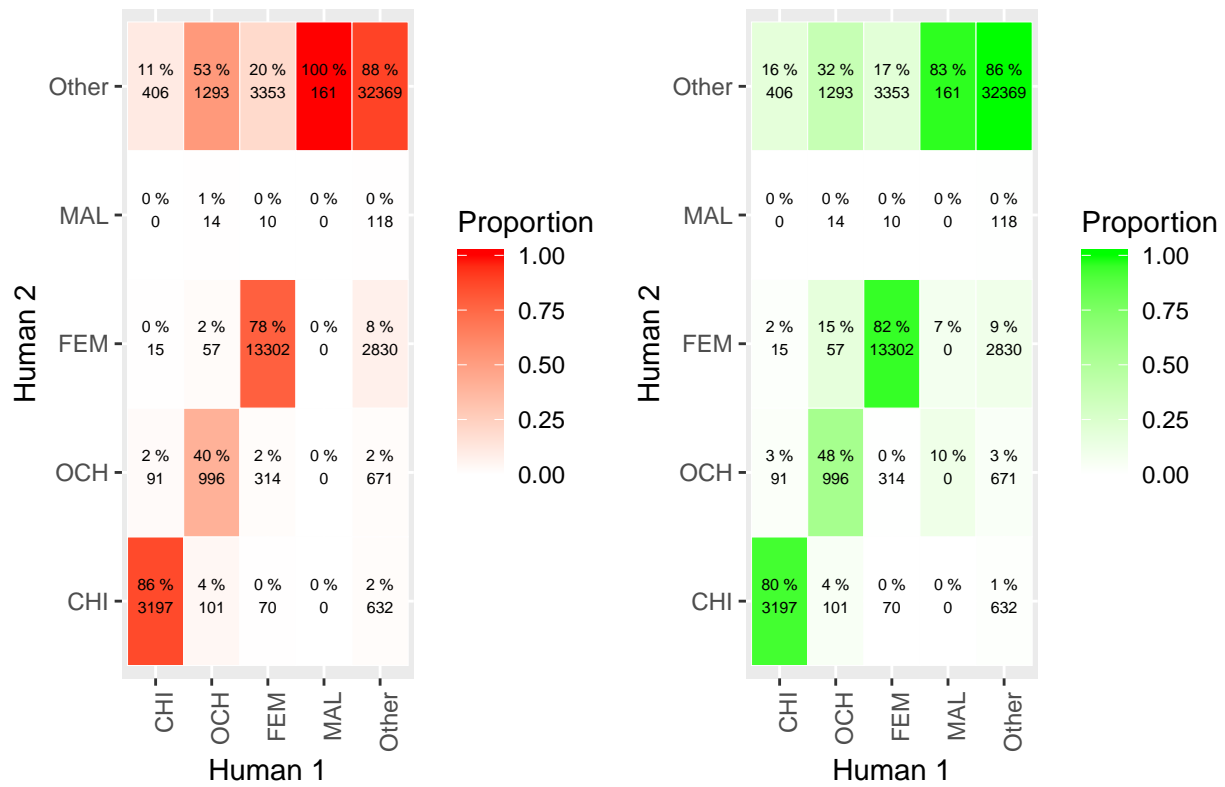
## Precision/Recall on TSE



## Precision/Recall on ROS



## Precision/Recall on SOD



## Cohen's kappa

As an other measure of the level of agreement between the two annotators, we propose the use of the Cohen's kappa measure, defined as follow :

$$\kappa = \frac{\Pr(\alpha) - \Pr(e)}{1 - \Pr(e)}$$

where:  $\Pr(\alpha)$  is the relative agreement between the annotators.  $\Pr(e)$  is the probability of a random agreement on a given frame.

If both annotators fully agree,  $\kappa = 1$ , if they fully disagree (or agree randomly),  $\kappa = 0$

| corpora | n_obs  | kappa     | weighted_kappa |
|---------|--------|-----------|----------------|
| WAR     | 60000  | 0.6883621 | 0.7634267      |
| SOD     | 60000  | 0.6844869 | 0.6899259      |
| ROW     | 60000  | 0.6661583 | 0.6879248      |
| all     | 360000 | 0.6403963 | 0.6531616      |
| ROS     | 60000  | 0.5707022 | 0.5808867      |
| BER     | 60000  | 0.6540080 | 0.5688731      |
| TSE     | 60000  | 0.5485063 | 0.5474113      |

## Vocalizations and turn-taking

For this study, we considered only the children that have been annotated as vcm or lex, for which vocalizations were classified as C, N, W, L, U, or Y. That led us to remove 7 children from the study, for a total of 53 children.

