# A bioinformatics solution to inter-rater agreement for forced time-alignment of data from underresourced languages

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## ANR-DFG DoReCo (2019-2022)

A collection of 50+ languages from documentations of small and often endangered languages.

- Morphological annotation (for 30+ languages).
- Phonemic alignment of all languages with MAUS.

#### **Evaluation of the MAUS alignment**

- 6 transcriptions (from 3 languages) for 4125 words
- Automatically aligned with MAUS (Kisler et al. 2017)
- Manually corrected by 4 annotators

## Languages

- Anal (Ozerov 2018) - 1341 word sample
- Sino-Tibetan language of north-east India
- Resígaro (Seifart 2009) – 657 word sample
- Arawakan language of Peru
- Vera'a (Schnell 2015) – 2127 word sample
- Austronesian language of Vanuatu

## Calculating inter-rater agreement

for annotations with different boundaries and different segmentation.

HI	bichar		vacţo	рс	1		hrangammamicha					
H2	bichar		vaeto	pe	1	hrang	2	amma		3 cha		
нз	bichar	$\langle \Box$	vacţo	рс	l hı		hrang <sub>m</sub> i <sub>a</sub> mma		2 cha			
H4	bichar	4	vacţo	pe		1 hrang	:	2 mi	3 amma <sup>4</sup>	cha		
66	95 Mov	ed bound	aries	Time (r)	Differen	t segmentation					68.37	

#### Figure 1: Correction of word alignment in Anal from four annotators (H1-H4)

Without knowing which unit corresponds to which, comparison is difficult. Current methods (Cohen 1968; Krippendorf 2004) don't match units between sequences:

- Atomization segmenting further into an equal number of constant segments
- relying on overlap and categorization (Holle & Rein 2015) EasyDIAg - using overlap to determine « nuclei » (Lücking et al. 2012)
- Staccato

One method does (Gamma, Mathet et al. 2015) with matching and agreement measurements in a single process; but it isn't easy to access

## The Needleman-Wunsch algorithm

An algorithm originally developed for matching DNA sequences (Needleman & Wunsch 1970).

Makes two sequences correspond by finding the alignments with the maximal number of matching units from all possible alignments

#### Implemented using the "pairwise2" function from Python's "Bio" library":

# pairwise2.align.globalms(annoA, annoB, 2, -1, -1, -1)

# Results

What the implementation returns is pairs / matches of same units from two different appetati

un	e sai	ne units nom t	w0 t	interent annota	nuons.							
113	144	145		146	147		148n			149	150	
114	144	145		146	147		148n	n	n	149	15	0
66	.92	Ma	ad hounds	, sien		Differen	it segmentation				68	1

Figure 2: Matching units of annotators H3-H4 (from figure 1, lines 3-4)

- Numbers indicate matched units; "n" unmatched units; numbers with "n" matched units with edited content
- Evaluation of the matching script by inter-rater agreement with a manual correction of the automatic matching result

#### Table 1: Score between automatic and manual unit matching

Language	Kappa-score	Accuracy				
Anal	0.98	98.25%				
Resigaro	0.9885	99.05%	•			
Vera'a	0.9727	97.6%				
Total	0.98	98%				

- Most disagreement is with pauses, The algorithm treats them like any other unit. Better results by
- removing / weighing?

#### Measuring the segmentation How to compare unit boundaries? Available measures for MAUS-human alignments (totals).

General unit matching:

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- 85,08% matched; 6,80% edited; 8,12% unmatched
- General overlap:
- 91,88% of matched units perfectly overlap Average distance of moved boundaries:

## 111ms for onsets; 133ms for offsets



# Other uses and perspectives

This algorithm can be used to align any sequences of behavioral data with different boundaries or mismatched segmentation (such as annotations of speech or gesture), to prepare sequences for measurement of inter-annotation agreement. It is also useful for aligning sequences that are coded at different levels, such as morphemes and words or words and utterances.

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