

Analysing translational survey data through multi-alignment

Michael Cysouw Jürg Fleischer

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Abstract

In the analysis of (dialectal) survey data there is a long path of interpretation from the data that is collected to the eventual interpretation. In most past (and current) research there is no paper-trail of all the large and small decisions being taken in the processing of the data. This paper describes a series of methods to document the processing of translational survey data, i.e. data that consists of translational equivalents. As an example we will process 2500 translations of a single sentence from the original Wenker data, transliterated from the original questionnaire from the 19th century. In addition to the well-known geographic distribution of sounds, we will show that it is just as well possible to extract syntactic and lexical variables from this data.

1 Multi-aligning translations

There are many ways in which comparable data can be collected to compare different language variants. Possibly the most traditional kind of comparable data (and also often criticised, REF?) are translational equivalent utterances. Language consultants are simply asked to produce the closest possible translation of a given utterance in their language. It is this kind of data that will be the focus of the paper, though the techniques proposed have a much wider application. We will analyse Wenker sentence 9: *‘Ich bin bei der Frau gewesen und ich habe es ihr gesagt, und sie sagte, sie wolle es auch ihrer Tochter sagen.’* (I have been at the women and I have told it to her, and she said that she would tell it to her daughter). We have transliterated about 2500 translations from the original Wenker questionnaire, extended with translations from Austria, Switzerland, the Netherlands, Belgium and various german-speaking linguistic enclaves. This data was used at the start of the 20th century to produce the infamous dialect maps.

[more]
[transliteration process]

2 Technical assumptions

- text based formats
- transparent manipulations
- keep detail as long as possible in the data processing pipeline

3 Formats for multi-alignment

3.1 Multi-alignment of sounds in words

Although one can argue that historical linguistics is from its inception based on a kind of multi-alignment, the first explicitly written-down multi-alignments arose in the context of DNA sequence analysis in biology (cf. Figure 1). In such files each element to be aligned (here amino-acids) are abbreviated with a single character, and the alignment is specified by fixing the position of the characters in a line. Typically, the first ten characters are reserved for a name, and afterwards the characters are just put directly after each other. When an element is missing for one of the species (as so-called ‘gap’), then a dash ‘-’ is inserted to keep the sequences aligned.

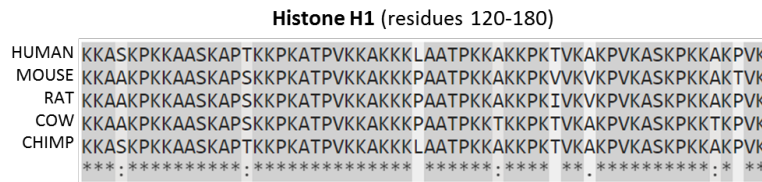


Figure 1: Example of a multiple sequence alignment of amino-acids across biological species.²

This approach can rather straightforwardly be adapted for the multiple alignment of sounds in comparative linguistics. An early example can be seen in Figure 2, aligning various words for ‘day’. However, one problem with linguistic data is that the assumption of ‘one element = one letter’ is difficult to maintain without strongly simplifying the data. For example, in phonetic transcriptions a single element often consists of multiple Unicode characters,

² Example from https://en.wikipedia.org/wiki/Sequence_alignment by T. Shafee.

like /a:/ or /tʃ/. In the benchmark data collected by List & Prokić (2014) they opted to put explicit tab marks between the columns, as illustrated by an example from their database in Figure 3.³ A recent development is to allow more columns with different kind of additional information in such files (e.g. data references or expert status). A multi-alignment then should be just a single column in such a file. Therefor, the column separator for multi-alignment has become a single space. This is found for example in the data underlying the article by Hill & List (2017), an excerpt of which is shown in Table 1.⁴ This approach is codified in the recent *Cross-Linguistic Data Formats* (CLDF) standard.⁵

D--E--N-	D--E--NY
Z--E--N-	DZ-E--N-
DZIE--N-	D--A--N-
DI-E--NA	D--E--IZ
D--I--A-	D--Y--DD
D--I--E-	Z-----U-
Z--U--E-	Z-----I-
J--O--UR	D-----I-
DJ-O--U-	G--IORNO

Figure 2: Linguistic multi-alignment from Bhargava & Kondrak (2009: 47).

ID	Form	Cognateset	Alignment	Source
4097	a ³¹ tsi ³¹	501	a ³¹ + ts i - - ³¹	Hill2017
2	pza ⁵⁵ pza ⁵⁵	33	p z a ⁵⁵ + p z a ⁵⁵	Hill2017
5462	pjã ²²	518	p j ã ~ ²²	Hill2017
9	pan ³¹ ʃɔʔ ⁵⁵	567	p a n ³¹ + ʃ ɔ ʔ ⁵⁵	Hill2017
10	pəŋ ³¹ ʃɔʔ ⁵⁵	567	p ə ŋ ³¹ + ʃ ɔ ʔ ⁵⁵	Hill2017

Table 1: Excerpt of data from Hill & List (2017)

There are two beta-stage software projects that try to develop a user interface to make it easier to manually edit linguistic multi-alignment: the Edictor

³ The full data is available online at <http://alignments.lingpy.org>

⁴ The full data is available at <https://zenodo.org/record/886179>.

⁵ The CLDF specification and documentation is available online at <http://cldf.clld.org>.

Germanic				
"all"				
American English.....	-	ɑ!	±	
Canadian English.....	-	ɒ!	±	
Central German (Cologne).....	?	a	l	
Central German (Honigberg).....	?	ɒ	±	
Central German (Luxembourg).....	?	a	l	
Central German (Murrhardt).....	?	a	l	
Danish.....	?	ɛ	l'	
Dutch (Antwerp).....	-	a	±	
Belgian Dutch.....	-	ɑ	±	
Dutch (Limburg).....	-	a	l	
Dutch (Ostend).....	-	ɑo	-	
Dutch.....	-	ɑ	±	

Figure 3: Linguistic multi-alignment with aligned multigraphs and using tabs as separators (List & Prokić 2014).

and the MSA-Editor.⁶ For the current example we have used the MSA-Editor to prepare the data.

3.2 multi-alignment of words in sentences

The concept of alignment between languages is also an important notion in the field of machine translation. Parallel translations between languages (or ‘bitexts’ as they are known in computational linguistics, see Tiedemann 2011) are the basic starting point for any machine translation, as this is the principal route to learn equivalent expressions between languages. The classical approach to learn equivalent structure between languages is to align bitexts, i.e. assign a linking between some parts of one sentence (‘words’) to parts of the translated sentence. The most influential software to perform such bitext alignment has been Giza++ (Och & Ney 2003). It used a specific output format for aligned sentences, which has become known as the ‘Pharaoh’ format.⁷ The Pharaoh-format counts the words, starting at zero, and then links these indices by a dash. For example, the bitext alignment shown in Figure 4 will be summarised in Pharaoh-format as ‘0-3 1-1 2-1 4-2’.

⁶ Both project have a common origin. The Editor is available at <https://github.com/digling/edictor>, described in List (2017). The MSA-Editor is available at <https://github.com/cysouw/msa-editor>.

⁷ Many names in the statistical machine translation have some kind of reference to Egypt, which seems to be a reference to one of the early highly influential machine translation toolkits, called ‘EGYPT’ <http://web.archive.org/web/20060826032742/http://www.clsp.jhu.edu:80/ws99/projects/mt/>.

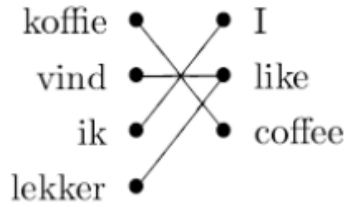


Figure 4: Bitext alignment (Tiedemann 2011: 4).

This Pharaoh-format cannot easily be generalised to a multi-alignment (i.e. for more than two languages). In machine translation there is not much use for alignments with more than two language, so no generalised format has arisen there. For that reason, the Cross-Linguistic Data Formats (CLDF) proposes a special format assigning abstract cluster-identifiers to each ‘column’ (maybe better called ‘virtual column’, or ‘functionalEquivalentset’ as they are called in the CLDF) in the multi-alignment.⁸

Form	Slice	Equivalentset
Koffie vind ik lekker	1	1
Koffie vind ik lekker	2	2
Koffie vind ik lekker	3	3
Koffie vind ik lekker	4	4
I like coffee	1	3
I like coffee	2	2
I like coffee	2	4
I like coffee	3	1
Ich mag gerne Kaffee	1	3
Ich mag gerne Kaffee	2	2
Ich mag gerne Kaffee	3	4
Ich mag gerne Kaffee	4	1

Table 2: multi-alignment in the format of the CLDF.

The obvious reason for this format is the crossing lines that occur regularly in word alignment. In sound alignment this also occurs (viz. in metathesis as shown in Figure 5), but it is not extremely common. For that reason, it is possible to use the ‘trick’ shown in Figure 5, namely to use two columns for the representation of one alignment.

⁸ See <https://github.com/cldf/cldf/blob/master/modules/ParallelText/README.md> for an explanation of the CLDF specification for parallel texts.

Language	Alignment							
Russian	s	ɔ	-	-	n	ts	ə	ej e nə
Polish	s	-	w	ɔ	n ^j	ts	ɛ	
French	s	ɔ	l	-	-	-	-	
Italian	s	o	l	-	-	-	-	
German	s	ɔ	-	-	-	-	-	
Swedish	s	u:	l	-	-	-	-	

Figure 5: Example of multi-alignment of metathesis (List 2014: 135).

In the general case this ‘virtual-column trick’ does not work for word alignments within sentences. However, when the languages compared are very similar (like with dialects) then it is possible to use this approach, which is more easily approachable to manual editing. This is the approach that we use for the current project. An excerpt of such an ‘virtual column’ alignment is shown in Table 3.

Sie	hat	gesagt	sie	will	es ₁	auch ₁	ihrer	Tochter	es ₂	auch ₂	sagen
si	hat	gesoat	si	wold	-	oah	hirrer	Doachter	ät	-	soan
sei	-	sot	sei	wollt	et	och	ihrer	Doochte	-	-	sohn
se	-	secht	se	wullt	es	-	ähri	Dochtr	-	ock	säga

Table 3: multi-alignment of dialect sentences using ‘virtual-column trick’

4 Workflow

All the alignments for this project were prepared manually. We have been testing various automatic approaches, but they turned out not to be helpful.⁹ The central problems for automatic approaches are that (i) the orthographic structure of our data is much too variable to pick up similarities and (ii) there is very little data for computational standards. Fortunately, the manual process is reasonably efficient by using our tool of choice, the web-based MSA-Editor. Also, the manual process helps identifying (and solving) the quirks in the data and makes the researcher intimately acquainted with the data, which turns

⁹ We tested an older version of the software package LingPy (List, Greenhill & Forkel 2017) and the standard biological software for multi-alignment ClustalX (Larkin et al. 2007). Standard word-alignment software like Giza++ (Och & Ney 2003) or fastalign (Dyer, Chahuneau & Smith 2013) need much more data to be feasible.

out to be highly useful for the later analysis and interpretation. In sum, the somewhat tedious, but manageable manual process is actually time well-spent with the data.

The workflow consists of three steps starting from the transliterated source. The results of each step are saved in different files, documenting the decisions being taken at each step. The resulting workflow (with the filenames used) are:

- transliterated source (file: 0source.tsv)
- multi-alignments of functionally equivalent words across the sentences (file: 1syntax.tsv)
- splitting up functionally equivalent words (i.e. the columns from the first step) into cognate sets (file: 2lexicon.tsv)
- multialigning the sounds within cognate words (file: 3sounds.tsv)

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