Forced Alignment of Spoken Audio

Josef Fruehwald 19 April 2016

Why Forced Alignment?

What we had

Data - Static

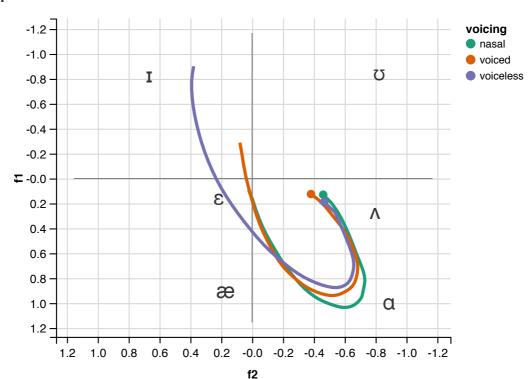


What we wanted:

Data - Dynamic



i:

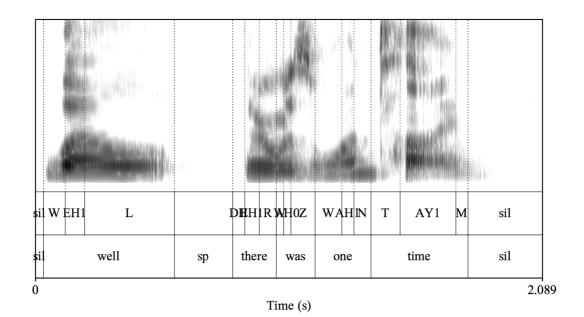


Getting from what we have to what we want

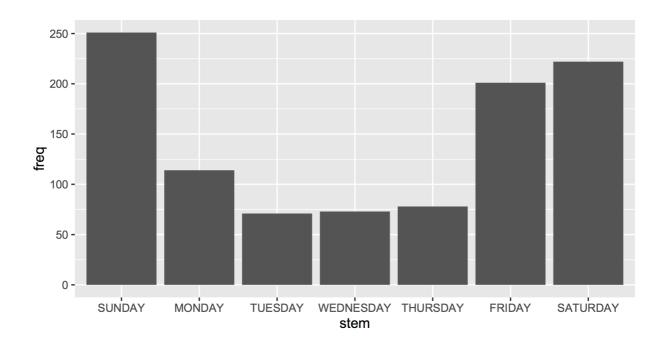
- Convert analogue recordings to digital format.
 - Preserve the most important metadata
- 2. Identify where in the audio speech sounds of interest are.
- 3. Automate the acoustic analysis of the speech sounds.
- 4. Apply statistical analysis to the acoustic analysis for inferences.

Identifying where in the audio speech sounds interest are.

"Forced Alignment""



Finding words in audio



Forced Alignment

Rest of the presentation:

- What some of the necessary bits and pieces are for doing forced alignment.
- What some of the tools out there are for doing alignment as an end user.

Bits and Pieces and Issues for doing forced alignment

Piece 1: A pronouncing dictionary

word pronunciation

well W EH1 L

there DH EH1 R

was W AHO Z

one W AH1 N

time T AY1 M

Issue 1: Pronunciation Variants

What do you do for multiple pronunciations? e.g. Bailey (2016)

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word pronunciation

walking W AO1 L K IHO N

walking W AO1 L K IHO NG

walking W AO1 L K IHO NG G
```

Issue 1: Pronunciation Variants

Option 1: Include all options

Let the aligner figure out which option to use.

Pros

You'll get more accurate timing.

· Cons

- In choosing pronunciation variants, some aligners have a lower rate of agreement with humans coders than humans coders do with each other (Bailey 2015)
- It can be tricky to identify which pronunciations are variants of each other.

Issue 1: Pronunciation Variants

Option 2: Only include one option
Only allow the aligner to choose one option

Pros

- It'll be easier to identify all instances of potential pronunciation variation.

· Cons

- The timing information will be less accurate.

Issue 2: Out of Dictionary Words

No matter how large a pronouncing dictionary you're working with, there will always be some words in free flowing speech that aren't in the dictionary.

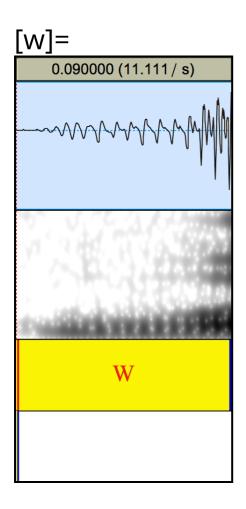
word pronunciation

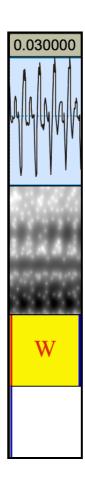
Fruehwald F R UW1 W AOO L D

hoagie HH OW1 G IY0

These either need to be added to the dictionary when the aligner is run, or a separate piece of software needs to try to guess the pronunciation based on the spelling.

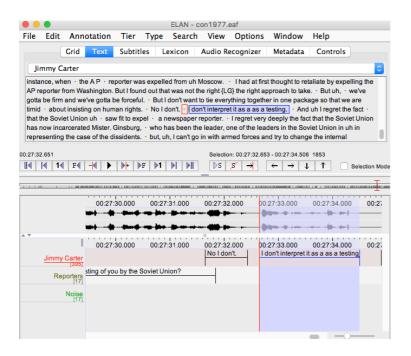
Piece 2: An acoustic model



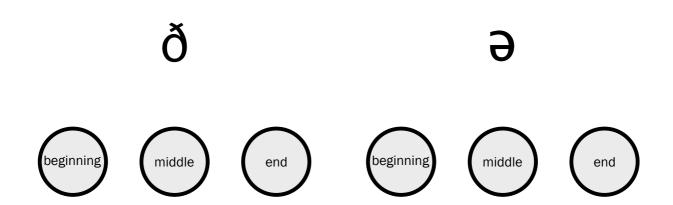


Piece 3: A transcript

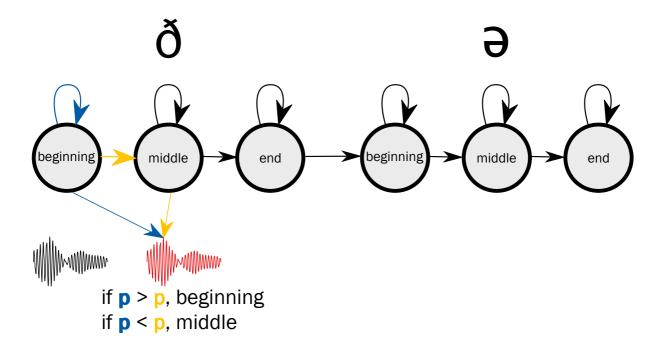
Outside of the original fieldwork, this is the most time consuming and expensive part.



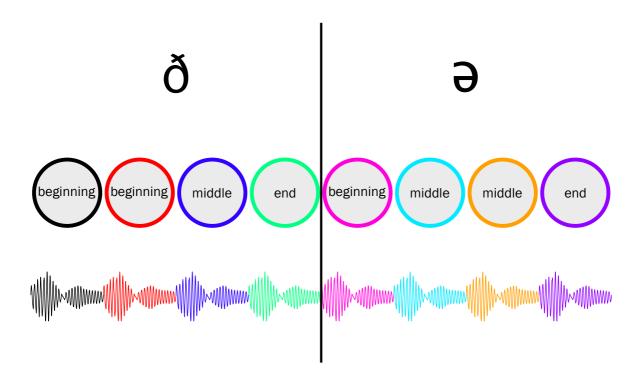
How it works



How it works



How it works



Concerns about forced alignment

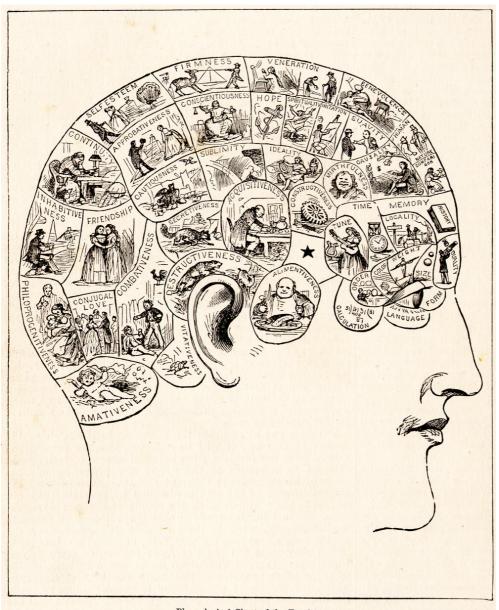
It'll make mistakes

- It is easier and faster (read: cheaper) to manually correct the output of automated systems than to create the annotations from scratch
- Humans make mistakes too! And the kinds of mistakes automated sytems make are usually systematic, so they're easier to identify and locate.

Concerns about forced alignment

It's a black box!

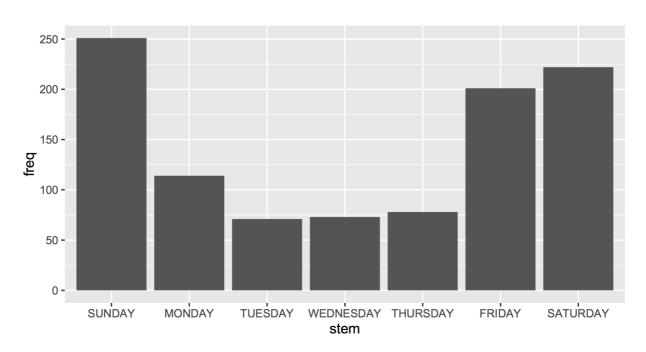
You are a black box



Phrenological Chart of the Faculties.

Concerns about forced alignment

Automation removes me from the data



Doing Forced Alignment at Home

FAVE

The FAVE-suite is actually two pieces of software: An aligner, and a Bayesian formant analyzer.

- Aligner based on p2fa, trained on 25 hours of US Supreme Court oral arguments.
- Fairly good time accuracy.

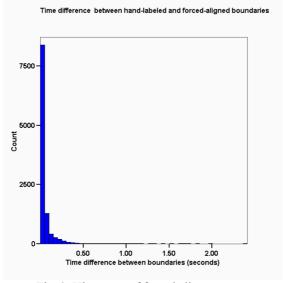


Fig. 3. Histogram of forced alignment errors.

FAVE Benefits

- Developed assuming that multiple talkers in the audio was the default case.
- Developed in the open, trying to be as crossplatform friendly as possible.
- Written in Python, which is a very widely understood programming language.
- The system is relatively simple and flexible (although its acoustic models are not).
- The primary developer is friendly and responsive

FAVE Cons

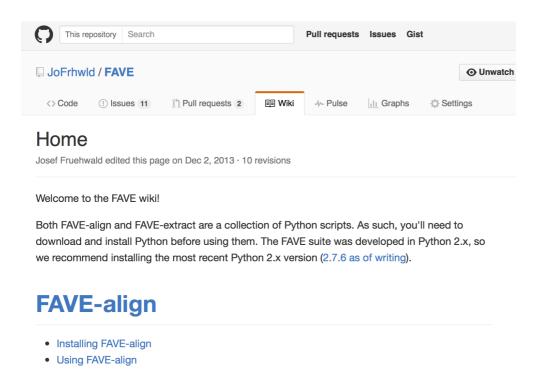
 Based on North American acoustic models, although MacKenzie & Turton have found it compares favorably to other aligners on British data.

	Median		Mean		Max	
	Onset	Offset	Onset	Offset	Onset	Offset
FAVE	0.009	0.009	0.019	0.021	0.583	0.588
PLA	0.015	0.019	0.267	0.252	55.473	55.488
SPPAS	0.150	0.155	0.504	0.480	68.903	67.408

Recommended FAVE Usage

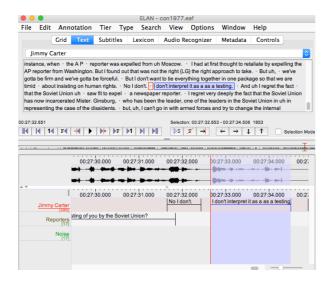
Download and install locally

Extensive documentation online, written assuming minimal familiarity with command line interfaces.



What FAVE needs as input

- Audio
- Transcriptions
 - Partially time aligned
 - Multiple speakers annotated separately



Prosodylab Aligner

Developed at University of McGill, Montreal

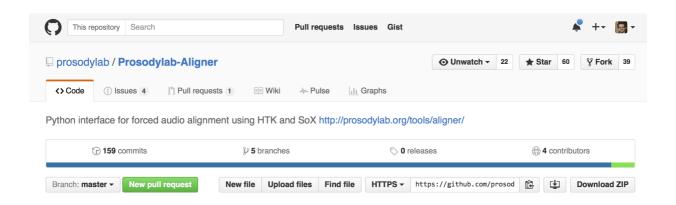
Pros & Cons

- Much the same as FAVE, but re-training of the acoustic models is built in.
- No streamlined facility yet for multiple talkers

Prosodylab Aligner

Recommended Usage

Download & Install



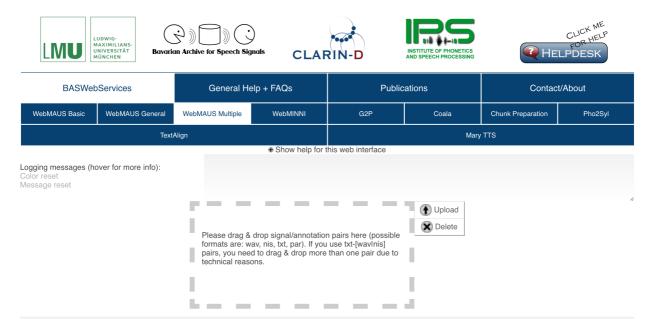
webMAUS

Developed in association with CLARIN-D

- Web-based platform
 - Easy to use
 - Less easy to adapt to task specific purposes
 - May be tricky if there are ethics restrictions on where and how your data is stored.
- No multiple talkers yet

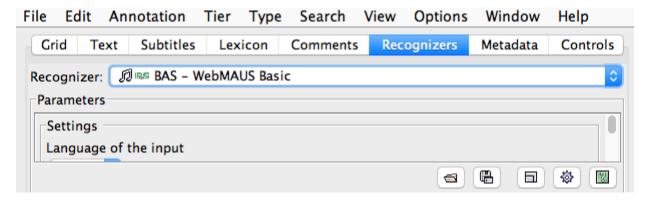
webMAUS

Recommended Usage



webMAUS

Recommended Usage



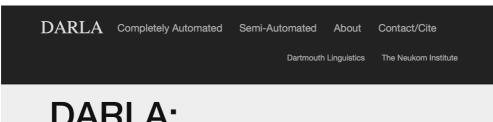
DARLA

System developed at Dartmouth University

- Pros
 - Includes an automatic speech recognition system.
- · Cons
 - So far, just a web-based service, with servers in the US

DARLA

Recommended Usage



DARLA: Dartmouth Linguistic Automation

A suite of vowel formant extraction tools tailored to research questions in sociophonetics.

Completely Automated Vowel Extraction

Audio → Formants

This automated method is designed for "big data" research projects where a general vowel space is prioritized over word-level transcription accuracy.

See details »

The End